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On some bounds on the perturbation of invariant subspaces of normal matrices with application to a graph connection problem

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Abstract

We provide upper bounds on the perturbation of invariant subspaces of normal matrices measured using a metric on the space of vector subspaces of \mathbb{C}^n in terms of the spectrum of both unperturbed and perturbed matrices as well as the spectrum of the unperturbed matrix only. The results presented give tighter bounds than the Davis–Kahan $\sin \Theta$ theorem. We apply the result to a graph perturbation problem.

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1 Introduction

Classical results on perturbation of invariant subspaces of a matrix usually take one of the two forms: (1) perturbation measured in terms of a natural metric in the space of vector subspaces (usually expressed as the sine of the angle between subspaces) with upper bound described in terms of the perturbation in the matrices as well as the spectra of both unperturbed and perturbed matrices (for example, the Davis–Kahan $\sin \Theta$ theorem [1] – see Section VIII.3 of [2] where a generalization of this theorem is given for normal matrices); or (2) perturbation measured in terms of bounds on norms of matrices that relate an invariant subspace with its perturbation in a more complex manner (which, in general, is not a natural metric in the space of vector subspaces) although the upper bound is based on the spectrum of the unperturbed matrix only (see, for example, [1, 3] or Chapter V of [4]).

In this paper¹ we first derive an upper bound reminiscent of the Davis–Kahan $\sin \Theta$ theorem, but generalized for normal matrices and with modestly tighter bound (Proposition 1). Then we use some geometric methods to derive a bound on perturbation measured in terms of a natural metric in the space of subspaces, but with upper bounds in terms of spectrum of the unperturbed matrix only (Proposition 2) when the spectrum is

¹A preprint version of this articles is posted on the arXiv preprint repository and can be accessed at <https://arxiv.org/abs/2103.09413> [5].

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well clustered (a relation formally described as “separation-preserving perturbation”). In the latter case our proposed result also allows easy identification of the perturbed invariant subspace (Lemma 7).

Definition 1 (Notations) Throughout the paper we assume $M, \tilde{M} \in \mathbb{C}^{n \times n}$ to be normal matrices unless specified otherwise, and by “eigenvectors” we refer to their right eigenvectors. The eigenvalues (not necessarily distinct) and corresponding unit eigenvectors (for degenerate eigenspaces, any orthonormal basis thereof) of M are λ_j and \mathbf{u}_j for $j = 1, 2, \dots, n$. Likewise, the eigenvalues and corresponding unit eigenvectors of \tilde{M} are $\tilde{\lambda}_j$ and $\tilde{\mathbf{u}}_j$ for $j = 1, 2, \dots, n$. We will usually consider the eigenvectors to be column vectors in $\mathbb{C}^{n \times 1}$. Let $U = [\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n]$ and $\tilde{U} = [\tilde{\mathbf{u}}_1, \tilde{\mathbf{u}}_2, \dots, \tilde{\mathbf{u}}_n]$ be the unitary matrices that diagonalize M and \tilde{M} respectively. A *dagger* as superscript on a matrix or a vector, $(\cdot)^\dagger$, denotes the conjugate transpose (Hermitian transpose) of the matrix or vector. For notational convenience, define $N = \{1, 2, \dots, n\}$.

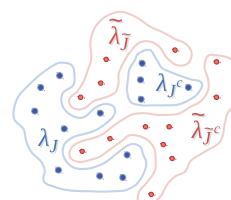
As a convention, we choose primed lower-case Latin letters to index variables (eigenvalues or eigenvectors) with tilde on them. Given a set $S \subseteq N$, we define the set $\mathbf{u}_S = \{\mathbf{u}_j | j \in S\}$. Likewise $\tilde{\mathbf{u}}_S = \{\tilde{\mathbf{u}}_{j'} | j' \in S\}$. Define the multi-sets $\lambda_S = \{\lambda_j | j \in S\}$ and $\tilde{\lambda}_S = \{\tilde{\lambda}_{j'} | j' \in S\}$ (by asserting that these are multi-sets, we allow multiplicity in the values, thus ensuring these sets have the same number of elements as S). We also define the *complement* of S as $S^c = N - S$.

The outline of the paper is as follows:

- 1 In Sect. 2.1 we describe a natural metric d_{sp} on $\text{Gr}(q, \mathbb{C}^n)$ (the space of q -dimensional complex vector subspaces of \mathbb{C}^n) to measure perturbation of invariant subspaces of $n \times n$ normal matrices. This metric is equivalent to the Frobenius norm of the $\sin \Theta$ matrix between subspaces of \mathbb{C}^n .
- 2 Some geometry lemmas are proven in Sect. 2.2, and then they are used in Sect. 3.3 for deriving bounds on the perturbation of invariant subspaces in terms of the spectrum of the unperturbed matrix only (when the spectrum is well clustered).
- 3 In Sect. 3.2 we describe an upper bound on the distance between invariant subspaces in terms of the spectrum of both unperturbed and perturbed matrices. Some of these results give improvements on the Davis–Kahan $\sin \Theta$ theorem for normal matrices (although the Davis–Kahan $\sin \Theta$ is usually stated for Hermitian matrices, there exist generalizations of the theorem for normal matrices – see Section VIII.3 of [2]). As an example (see Fig. 1), for any $J, \tilde{J} \subseteq N$, with $|J| = |\tilde{J}| = q$, Proposition 1 states

$$d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) \leq \sqrt{\frac{1}{q} \sum_{j \in J} \frac{\|(\tilde{M} - M)\mathbf{u}_j\|_2^2 - \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}}$$

Figure 1 Partition of the eigenvalues of M (in blue) and \tilde{M} (in red)



with

$$\kappa_j = \begin{cases} 0, & \text{if } \|(\tilde{M} - M)\mathbf{u}_j\|_2 \geq \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|, \\ 1, & \text{if } \|(\tilde{M} - M)\mathbf{u}_j\|_2 < \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|. \end{cases}$$

This is a tighter upper bound than the Davis–Kahan $\sin \Theta$ theorem, which, as a consequence, leads to the rediscovery of a couple of slight variations on the Davis–Kahan $\sin \Theta$ theorem in Corollary 5, where, as an example, one result states

$$d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) \leq \frac{\min(1, \sqrt{\frac{n-q}{q}})}{\max(\text{sep}(\lambda_J, \tilde{\lambda}_{\tilde{J}^c}), \text{sep}(\lambda_{J^c}, \tilde{\lambda}_{\tilde{J}}))} \|\tilde{M} - M\|_2,$$

where $\text{sep}(P, Q) = \min_{p \in P, q \in Q} |p - q|$ simply measures the min-min distance between the sets (this is unlike the Davis–Kahan $\sin \Theta$ theorem generalized for normal matrices, where it is necessary to find a ‘strip’ or ‘annulus’ of width δ separating λ_J and $\tilde{\lambda}_{\tilde{J}^c}$ – see Theorem VIII.3.1 of [2]).

- 4 The next set of the main results of this paper appears in Sect. 3.3, which formalizes the notion of well-clustered spectrum in Lemma 7, followed by Proposition 2 that provides the upper bound on the perturbation of an invariant subspace in terms of the spectrum of the unperturbed matrix only. These results rely on the geometry lemmas from Sect. 2.2. As an example, one of the results of Proposition 2 states that if $\|\tilde{M} - M\|_2 < \frac{1}{2} \text{sep}(\lambda_J, \lambda_{J^c})$, then

$$\begin{aligned} d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) &\leq \frac{1}{\sqrt{q}} \min \left(\sqrt{\sum_{j \in J} \left(\frac{\|(\tilde{M} - M)\mathbf{u}_j\|_2}{\min_{k \in J^c} |\lambda_k - \lambda_j| - \|\tilde{M} - M\|_2} \right)^2}, \right. \\ &\quad \left. \sqrt{\sum_{j \in J^c} \left(\frac{\|(\tilde{M} - M)\mathbf{u}_j\|_2}{\min_{k \in J} |\lambda_k - \lambda_j| - \|\tilde{M} - M\|_2} \right)^2} \right) \\ &\leq \min \left(1, \sqrt{\frac{n-q}{q}} \right) \frac{\|\tilde{M} - M\|_2}{\text{sep}(\lambda_J, \lambda_{J^c}) - \|\tilde{M} - M\|_2}, \end{aligned}$$

where $\hat{J} = \{j' \mid \min_{j \in N} |\tilde{\lambda}_{j'} - \lambda_j| = \min_{j \in J} |\tilde{\lambda}_{j'} - \lambda_j|\}$ is the set of indices corresponding to the eigenvalues of \tilde{M} that are closer to λ_J than to λ_{J^c} .

- 5 Sect. 4 demonstrates an application to the perturbation of a null-space of a matrix in the context of a graph perturbation problem.

2 Preliminaries

2.1 A metric on $\text{Gr}(q, \mathbb{C}^n)$

Definition 2 (Subspace distance) Suppose that $X, Y \subseteq \mathbb{C}^n$ are q -dimensional vector subspaces of \mathbb{C}^n .

Let $\{\mathbf{x}_j\}_{j=1,2,\dots,q}$ and $\{\mathbf{y}_j\}_{j=1,2,\dots,q}$ be orthonormal bases on X and Y . The subspace distance between X and Y is defined as

$$d_{\text{sp}}(X, Y) = \frac{1}{\sqrt{2q}} \|\mathbf{X}\mathbf{X}^\dagger - \mathbf{Y}\mathbf{Y}^\dagger\|_F, \quad (1)$$

where

$$\mathbf{X} = [\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_q] \quad \text{and} \quad \mathbf{Y} = [\mathbf{y}_1, \mathbf{y}_2, \dots, \mathbf{y}_q] \quad (2)$$

are the $n \times q$ matrices in which the columns represent the unit vectors $\{\mathbf{x}_j\}_{j=1,2,\dots,q}$ and $\{\mathbf{y}_j\}_{j=1,2,\dots,q}$. \mathbf{X}^\dagger and \mathbf{Y}^\dagger are the Hermitian transpose (*i.e.*, adjoint) of \mathbf{X} and \mathbf{Y} respectively.

Note that the matrices $\mathbf{X}\mathbf{X}^\dagger$ and $\mathbf{Y}\mathbf{Y}^\dagger$ are the projection operators on X and Y respectively. The space of difference of such projection operators is well studied in the literature (see [6, 7] for example), and the norms of such differences have been used as a metric on $\text{Gr}(q, \mathbb{C}^n)$ (see [8] for example). In fact this metric is equivalent to the Frobenius norm of the $\sin \Theta$ matrix between subspaces of \mathbb{C}^n that is used for measuring perturbation of invariant subspaces in the context of the Davis–Kahan $\sin \Theta$ theorem. We choose the Frobenius norm for measuring the distance between the projection operators and use a scaling factor of $\frac{1}{\sqrt{2q}}$ for convenience and some additional properties of the metric. The following lemmas outline some elementary and mostly standard properties of this metric.

Let X^\perp and Y^\perp be orthogonal complements of X and Y respectively in \mathbb{C}^n . Let $\{\mathbf{x}_j\}_{j=q+1,q+2,\dots,n}$ and $\{\mathbf{y}_k\}_{k=q+1,q+2,\dots,n}$ be orthonormal basis for X^\perp and Y^\perp respectively. Define

$$\mathbf{X}^\perp = [\mathbf{x}_{q+1}, \mathbf{x}_{q+2}, \dots, \mathbf{x}_n] \quad \text{and} \quad \mathbf{Y}^\perp = [\mathbf{y}_{q+1}, \mathbf{y}_{q+2}, \dots, \mathbf{y}_n]. \quad (3)$$

Lemma 1 (Equivalent forms of d_{sp})

$$\begin{aligned} 1 \quad d_{\text{sp}}(X, Y) &= \sqrt{1 - \frac{1}{q} \|\mathbf{X}^\dagger \mathbf{Y}\|_F^2} = \sqrt{1 - \frac{1}{q} \sum_{j=1}^q \sum_{k=1}^q |\mathbf{x}_j^\dagger \mathbf{y}_k|^2} \\ 2 \quad d_{\text{sp}}(X, Y) &= \sqrt{\frac{1}{q} \|\mathbf{X}^\perp \mathbf{Y}\|_F^2} = \sqrt{\frac{1}{q} \sum_{j=q+1}^n \sum_{k=1}^q |\mathbf{x}_j^\dagger \mathbf{y}_k|^2} \end{aligned}$$

Proof

- 1 In the following we use the definition $\|\mathbf{A}\|_F^2 = \text{tr}(\mathbf{A}^\dagger \mathbf{A})$ and the property that $\text{tr}(\mathbf{AB}) = \text{tr}(\mathbf{BA})$.

$$\begin{aligned} & (d_{\text{sp}}(X, Y))^2 \\ &= \frac{1}{2q} \|\mathbf{X}\mathbf{X}^\dagger - \mathbf{Y}\mathbf{Y}^\dagger\|_F^2 \\ &= \frac{1}{2q} \text{tr}((\mathbf{X}\mathbf{X}^\dagger - \mathbf{Y}\mathbf{Y}^\dagger)^\dagger (\mathbf{X}\mathbf{X}^\dagger - \mathbf{Y}\mathbf{Y}^\dagger)) \\ &= \frac{1}{2q} \text{tr}(\mathbf{X}\mathbf{X}^\dagger \mathbf{X}\mathbf{X}^\dagger) + \text{tr}(\mathbf{Y}\mathbf{Y}^\dagger \mathbf{Y}\mathbf{Y}^\dagger) - \text{tr}(\mathbf{X}\mathbf{X}^\dagger \mathbf{Y}\mathbf{Y}^\dagger) - \text{tr}(\mathbf{Y}\mathbf{Y}^\dagger \mathbf{X}\mathbf{X}^\dagger) \\ &= \frac{1}{2q} (\text{tr}(\mathbf{X}\mathbf{X}^\dagger) + \text{tr}(\mathbf{Y}\mathbf{Y}^\dagger) - 2 \text{tr}(\mathbf{Y}^\dagger \mathbf{X}\mathbf{X}^\dagger \mathbf{Y})) \\ & \quad (\text{since } \mathbf{X}^\dagger \mathbf{X} = \mathbf{Y}^\dagger \mathbf{Y} = I) \\ &= 1 - \frac{1}{q} \|\mathbf{X}^\dagger \mathbf{Y}\|_F^2 \\ & \quad \left(\text{since } \text{tr}(\mathbf{X}\mathbf{X}^\dagger) = \text{tr}(\mathbf{X}^\dagger \mathbf{X}) = \sum_{j=1}^q \mathbf{x}_j^\dagger \mathbf{x}_j = q, \text{ and likewise for } \mathbf{Y} \right) \\ &= 1 - \frac{1}{q} \sum_{j=1}^q \sum_{k=1}^q |\mathbf{x}_j^\dagger \mathbf{y}_k|^2. \end{aligned}$$

- 2 Note that $[\mathbf{X}, \mathbf{X}^\perp]$ is an $n \times n$ unitary matrix with columns being the vectors of the orthonormal basis $\{\mathbf{x}_i\}_{i=1,2,\dots,n}$. Thus, $[\mathbf{X}, \mathbf{X}^\perp][\mathbf{X}, \mathbf{X}^\perp]^\dagger = \mathbf{X}\mathbf{X}^\dagger + \mathbf{X}^\perp\mathbf{X}^{\perp\dagger} = I$. Thus,

$$\begin{aligned}(d_{\text{sp}}(X, Y))^2 &= 1 - \frac{1}{q} \|\mathbf{X}^\dagger \mathbf{Y}\|_F^2 \\&= 1 - \frac{1}{q} \text{tr}(\mathbf{Y}^\dagger \mathbf{X} \mathbf{X}^\dagger \mathbf{Y}) \\&= 1 - \frac{1}{q} \text{tr}(\mathbf{Y}^\dagger (I - \mathbf{X}^\perp \mathbf{X}^{\perp\dagger}) \mathbf{Y}) \\&= 1 - \frac{1}{q} \text{tr}(\mathbf{Y}^\dagger \mathbf{Y}) + \frac{1}{q} \text{tr}(\mathbf{Y}^\dagger \mathbf{X}^\perp \mathbf{X}^{\perp\dagger} \mathbf{Y}) \\&= 1 - \frac{1}{q} + \frac{1}{q} \text{tr}(\mathbf{Y}^\dagger \mathbf{X}^\perp \mathbf{X}^{\perp\dagger} \mathbf{Y}) \\&= \frac{1}{q} \|\mathbf{X}^{\perp\dagger} \mathbf{Y}\|_F^2 = \frac{1}{q} \sum_{j=q+1}^n \sum_{k=1}^q |\mathbf{x}_j^\dagger \mathbf{y}_k|^2.\end{aligned}$$

□

Lemma 2 (Properties of d_{sp})

- 1 The value of $d_{\text{sp}}(X, Y)$ is independent of the choice of basis on X or Y (or the basis on X^\perp or Y^\perp , if using the equivalent form in Lemma 1.2).
- 2 d_{sp} is a metric on $\text{Gr}(q, \mathbb{C}^n)$ (the space of q -dimensional complex subspaces of \mathbb{C}^n).
- 3 $\sqrt{q} d_{\text{sp}}(X, Y) = \sqrt{n-q} d_{\text{sp}}(X^\perp, Y^\perp)$.
- 4 $d_{\text{sp}}(X, Y) \leq 1$, with equality holding iff X and Y are orthogonal subspaces (which is possible only if $q \leq n/2$).

Proof

- 1 Suppose that $\{\mathbf{x}'_j\}_{j=1,2,\dots,q}$ and $\{\mathbf{y}'_j\}_{j=1,2,\dots,q}$ are a different set of orthonormal bases on X and Y respectively. Define $\mathbf{X}' = [\mathbf{x}'_1, \mathbf{x}'_2, \dots, \mathbf{x}'_q]$, $\mathbf{Y}' = [\mathbf{y}'_1, \mathbf{y}'_2, \dots, \mathbf{y}'_q]$. Thus there exist $q \times q$ unitary matrices $R_X, R_Y \in U(q)$ such that $\mathbf{X} = \mathbf{X}' R_X$ and $\mathbf{Y} = \mathbf{Y}' R_Y$. Then

$$\begin{aligned}(d_{\text{sp}}(X, Y))^2 &= \frac{1}{2q} \|\mathbf{X} \mathbf{X}^\dagger - \mathbf{Y} \mathbf{Y}^\dagger\|_F^2 \\&= \frac{1}{2q} \|(\mathbf{X}' R_X)(\mathbf{X}' R_X)^\dagger - (\mathbf{Y}' R_Y)(\mathbf{Y}' R_Y)^\dagger\|_F^2 \\&= \frac{1}{2q} \|\mathbf{X}' \mathbf{X}'^\dagger - \mathbf{Y}' \mathbf{Y}'^\dagger\|_F^2.\end{aligned}$$

For the equivalent form in Lemma 1.2 we can use the orthonormal basis $\{\mathbf{x}'_j\}_{j=q+1,q+2,\dots,n}$ and $\{\mathbf{y}'_k\}_{k=q+1,q+2,\dots,n}$ for X^\perp and Y^\perp respectively and analogously derive at the equivalent form using the primed basis.

- 2 *Nonnegativity* and *symmetry* properties are obvious from the definition of d_{sp} .

If X and Y are the same subspaces, we can choose the same basis for them (since the value of $d_{\text{sp}}(X, Y)$ is independent of the choice of a basis on X and Y), doing so makes it obvious that $d_{\text{sp}}(X, Y) = 0$.

Triangle inequality holds due to the fact that Frobenius norm of the difference of matrices is a metric on $\mathbb{C}^{n \times n}$.

- 3 Note that X^\perp and Y^\perp are $(n - q)$ -dimensional subspaces of \mathbb{C}^n . Furthermore, X is the orthogonal complement of X^\perp . As a consequence, due to Lemma 1.2,

$$\begin{aligned} d_{\text{sp}}(X^\perp, Y^\perp) &= \sqrt{\frac{1}{n-q}} \|\mathbf{X}^\dagger \mathbf{Y}^\perp\|_F \\ &= \sqrt{\frac{1}{n-q}} \|\mathbf{Y}^{\perp\dagger} \mathbf{X}\|_F \quad (\text{since } \|\mathbf{A}\|_F = \|\mathbf{A}^\dagger\|_{F^*}) \\ &= \sqrt{\frac{1}{n-q}} \sqrt{q} d_{\text{sp}}(Y, X) = \sqrt{\frac{q}{n-q}} d_{\text{sp}}(X, Y). \end{aligned}$$

□

- 4 The last property is obvious from the result of Lemma 1.1.

2.2 Some results involving set distances

In this section we provide some geometry results that will be used in Sect. 3.3 for computing the upper bounds on the perturbation of invariant subspaces in terms of the spectrum of the unperturbed matrix only. For the purpose of this paper and for simplicity, we consider only closed subsets of metric spaces in the following lemmas, although all these results can potentially be generalized for subsets that are open or/and closed in the metric space.

Definition 3 Given closed subsets A, B of a metric space (Ψ, d) , we define the following:

- 1 *Separation between the sets*

$$\text{sep}(A, B) = \min_{\substack{a \in A, \\ b \in B}} d(a, b);$$

- 2 *Hausdorff distance between the sets*

$$d_H(A, B) = \max \left(\max_{a \in A} \min_{b \in B} d(a, b), \max_{b \in B} \min_{a \in A} d(a, b) \right);$$

- 3 *Diameter of a set*

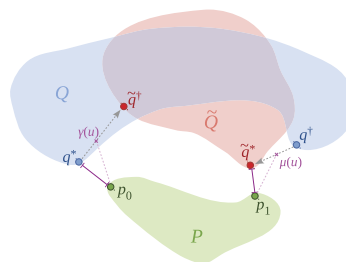
$$\text{diam}(A) = \max_{\substack{a \in A, \\ a' \in A}} d(a, a').$$

Lemma 3 If (Ψ, d) is a metric space, then for any closed subsets $P, Q, R \subseteq \Psi$,

$$\text{sep}(P, Q) \leq \text{sep}(P, R) + \text{sep}(R, Q) + \text{diam}(R). \quad (4)$$

Proof Let $(p^*, r_1) \in \arg \min_{p \in P, r \in R} d(p, r)$ (that is, $p^* \in P, r_1 \in R$ are a pair of points such that $d(p^*, r_1) = \min_{p \in P, r \in R} d(p, r) = \text{sep}(P, R)$). Likewise, let $(q^*, r_2) \in \arg \min_{q \in Q, r \in R} d(q, r)$ (that is, $d(q^*, r_2) = \text{sep}(R, Q)$). Then

$$\begin{aligned} \text{sep}(P, Q) &\leq d(p^*, q^*) \quad \left(\text{since } \text{sep}(P, Q) = \min_{\substack{p \in P, \\ q \in Q}} d(p, q) \right) \\ &\leq d(p^*, r_1) + d(r_1, q^*) \quad (\text{triangle inequality}) \\ &= \text{sep}(P, R) + d(r_1, q^*) \end{aligned}$$

Figure 2 Illustration for the proof of Lemma 4


$$\begin{aligned}
 &\leq \text{sep}(P, R) + d(r_1, r_2) + d(q^*, r_2) \quad (\text{triangle inequality}) \\
 &= \text{sep}(P, R) + \text{sep}(R, Q) + d(r_1, r_2) \\
 &\leq \text{sep}(P, R) + \text{sep}(R, Q) + \text{diam}(R).
 \end{aligned} \tag{5}$$

□

Lemma 4 If (Ψ, d) is a connected path metric space, then for any closed subsets $P, Q, \tilde{Q} \subseteq \Psi$,

$$\text{sep}(P, Q) \leq \text{sep}(P, \tilde{Q}) + d_H(\tilde{Q}, Q). \tag{6}$$

Proof Let $(p_0, q^*) \in \arg \min_{p \in P, q \in Q} d(p, q)$ (that is, $p_0 \in P, q^* \in Q$ are a pair of points such that $d(p_0, q^*) = \min_{p \in P, q \in Q} d(p, q)$) – see Fig. 2. Likewise, let $(p_1, \tilde{q}^*) \in \arg \min_{p \in P, q' \in \tilde{Q}} d(p, q')$. Furthermore, let $\tilde{q}^\dagger \in \arg \min_{q' \in \tilde{Q}} d(q^*, q')$ and $q^\dagger \in \arg \min_{q \in Q} d(q, \tilde{q}^*)$.

Consider the shortest path $\gamma : [0, 1] \rightarrow \Psi$ connecting q^* and \tilde{q}^\dagger and parameterized by the normalized distance from q^* , so that $\gamma(0) = q^*, \gamma(1) = \tilde{q}^\dagger$ and

$$d(q^*, \gamma(u)) = ud(q^*, \tilde{q}^\dagger). \tag{7}$$

Likewise, $\mu : [0, 1] \rightarrow \Psi$ be the shortest path connecting q^\dagger and \tilde{q}^* , and parameterized by the normalized distance from q^\dagger , so that $\mu(0) = q^\dagger, \mu(1) = \tilde{q}^*$ and $d(q^\dagger, \mu(u)) = ud(q^\dagger, \tilde{q}^*)$. Consequently, since $\mu(u)$ is a point on the shortest path connecting q^\dagger and \tilde{q}^* , we have

$$d(\mu(u), \tilde{q}^*) = d(q^\dagger, \tilde{q}^*) - d(q^\dagger, \mu(u)) = (1 - u)d(q^\dagger, \tilde{q}^*). \tag{8}$$

Define $f : [0, 1] \rightarrow \mathbb{R}$ as $f(t) = d(p_0, \gamma(t))$, and $g : [0, 1] \rightarrow \mathbb{R}$ as $g(t) = d(p_1, \mu(t))$. It is easy to note that both f and g are continuous.

As a consequence, we have the following:

$$\begin{aligned}
 f(0) &= d(p_0, q^*) = \min_{\substack{p \in P, \\ q \in Q}} d(p, q) \leq d(p_1, q^\dagger) = g(0), \\
 g(1) &= d(p_1, \tilde{q}^*) = \min_{\substack{p \in P, \\ q' \in \tilde{Q}}} d(p, q') \leq d(p_0, \tilde{q}^\dagger) = f(1).
 \end{aligned}$$

Thus, by intermediate value theorem, there exists $u \in [0, 1]$ such that $f(u) = g(u)$. That is,

$$d(p_0, \gamma(u)) = d(p_1, \mu(u)) \quad \text{for some } u \in [0, 1]. \tag{9}$$

Using this, we have

$$\begin{aligned}
 \min_{\substack{p \in P, \\ q \in Q}} d(p, q) &= d(p_0, q^*) \\
 &\leq d(p_0, \gamma(u)) + d(q^*, \gamma(u)) \quad (\text{triangle inequality}) \\
 &= d(p_1, \mu(u)) + d(q^*, \gamma(u)) \quad (\text{using (9)}) \\
 &\leq d(p_1, \tilde{q}^*) + d(\mu(u), \tilde{q}^*) + d(q^*, \gamma(u)) \quad (\text{triangle inequality}) \\
 &= \min_{\substack{p \in P, \\ q' \in \tilde{Q}}} d(p, q') + d(\mu(u), \tilde{q}^*) + d(q^*, \gamma(u)) \\
 &= \min_{\substack{p \in P, \\ q' \in \tilde{Q}}} d(p, q') + (1-u)d(q^\dagger, \tilde{q}^*) + ud(q^*, \tilde{q}^\dagger) \quad (\text{using (7) and (8)}) \\
 &\leq \min_{\substack{p \in P, \\ q' \in \tilde{Q}}} d(p, q') + \max(d(q^\dagger, \tilde{q}^*), d(q^*, \tilde{q}^\dagger)) \\
 &= \min_{\substack{p \in P, \\ q' \in \tilde{Q}}} d(p, q') + \max\left(\min_{q \in Q} d(q, \tilde{q}^*), \min_{q' \in \tilde{Q}} d(q^*, q')\right) \\
 &\quad (\text{definitions of } q^\dagger \text{ and } \tilde{q}^\dagger) \\
 &\leq \min_{\substack{p \in P, \\ q' \in \tilde{Q}}} d(p, q') + \max\left(\max_{q' \in \tilde{Q}} \min_{q \in Q} d(q, q'), \max_{q \in Q} \min_{q' \in \tilde{Q}} d(q, q')\right) \\
 &= \text{sep}(P, \tilde{Q}) + d_H(\tilde{Q}, Q).
 \end{aligned}$$

□

Lemma 5 Suppose that P, Q, \tilde{R} are closed subsets of a metric space (Ψ, d) such that

$$\max_{r' \in \tilde{R}} \min_{s \in P \cup Q} d(s, r') + d_H(P \cup Q, \tilde{R}) < \text{sep}(P, Q) \quad (10)$$

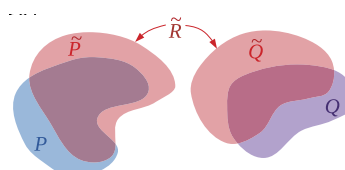
Define (see Fig. 3) $\tilde{P}, \tilde{Q} \subseteq \tilde{R}$ such that

$$\begin{aligned}
 \tilde{P} &= \left\{ r' \in \tilde{R} \mid \min_{s \in P \cup Q} d(s, r') = \min_{p \in P} d(p, r') \right\} \quad \text{and} \\
 \tilde{Q} &= \left\{ r' \in \tilde{R} \mid \min_{s \in P \cup Q} d(s, r') = \min_{q \in Q} d(q, r') \right\}.
 \end{aligned} \quad (11)$$

Then

- 1 $\{\tilde{P}, \tilde{Q}\}$ constitutes a partition of \tilde{R} .
- 2 $\arg \min_{s \in P \cup Q} d(s, p') \subseteq P, \forall p' \in \tilde{P}$, and $\arg \min_{s \in P \cup Q} d(s, q') \subseteq Q, \forall q' \in \tilde{Q}$.
(Consequently, $\min_{s \in P \cup Q} d(s, p') = \min_{s \in P} d(s, p'), \forall p' \in \tilde{P}$, and $\min_{s \in P \cup Q} d(s, q') = \min_{s \in Q} d(s, q'), \forall q' \in \tilde{Q}$.)

Figure 3 Illustration for Lemma 5



- 3 $\arg \min_{r' \in \tilde{R}} d(p, r') \subseteq \tilde{P}, \forall p \in P$, and $\arg \min_{r' \in \tilde{R}} d(q, r') \subseteq \tilde{Q}, \forall q \in Q$. (Consequently,
 $\min_{r' \in \tilde{R}} d(p, r') = \min_{r' \in \tilde{P}} d(p, r'), \forall p \in P$, and $\min_{r' \in \tilde{R}} d(q, r') = \min_{r' \in \tilde{Q}} d(q, r'), \forall q \in Q$.)
- 4 $d_H(P, \tilde{P}) \leq d_H(P \cup Q, \tilde{R}), d_H(Q, \tilde{Q}) \leq d_H(P \cup Q, \tilde{R})$, and
 $\max(d_H(P, \tilde{P}), d_H(Q, \tilde{Q})) = d_H(P \cup Q, \tilde{R})$.
- 5 If (Ψ, d) is a connected path metric space, then $\text{sep}(\tilde{P}, \tilde{Q}) \geq \text{sep}(P, Q) - 2d_H(P \cup Q, \tilde{R})$.
- If the above holds, we say “ \tilde{R} is a separation-preserving perturbation of P and Q ” and call $\{\tilde{P}, \tilde{Q}\}$ to be the “separation-preserving partition of \tilde{R} ”.

Proof 1. We first prove that $\{\tilde{P}, \tilde{Q}\}$ constitutes a partition of \tilde{R} .

Proof for $\tilde{P} \cup \tilde{Q} = \tilde{R}$: For fixed $r' \in \tilde{R}$, an element of $\arg \min_{s \in P \cup Q} d(s, r')$ is either in P or in Q . In the former case the point r' will belong to \tilde{P} , while in the latter case it will belong to \tilde{Q} (with the possibility that it belongs to both) due to Definition (11). Thus there does not exist a point $r' \in \tilde{R}$ that does not belong to either \tilde{P} or \tilde{Q} .

Proof for $\tilde{P} \cap \tilde{Q} = \emptyset$: We prove this by contradiction. If possible, let $\rho' \in \tilde{P} \cap \tilde{Q}$. Since $\rho' \in \tilde{P}$, due to Definition (11), there exists $p_1 \in P$ such that $\min_{s \in P \cup Q} d(s, \rho') = d(p_1, \rho')$. Likewise, there exists $q_1 \in Q$ such that $\min_{s \in P \cup Q} d(s, \rho') = d(q_1, \rho')$. Thus,

$$\begin{aligned}
 2 \min_{s \in P \cup Q} d(s, \rho') &= d(p_1, \rho') + d(q_1, \rho') \\
 &\geq d(p_1, q_1) \quad (\text{triangle inequality}) \\
 &\geq \min_{\substack{p \in P, \\ q \in Q}} d(p, q) \quad (\text{since } p_1 \in P, q_1 \in Q) \\
 \Rightarrow 2 \max_{r' \in \tilde{R}} \min_{s \in P \cup Q} d(s, r') &\geq \min_{\substack{p \in P, \\ q \in Q}} d(p, q) \\
 \Rightarrow \max_{r' \in \tilde{R}} \min_{s \in P \cup Q} d(s, r') + d_H(P \cup Q, \tilde{R}) &\geq \text{sep}(P, Q).
 \end{aligned}$$

This contradicts assumption (10) of the lemma. Hence there cannot exist a $\rho' \in \tilde{P} \cap \tilde{Q}$. Thus $\tilde{P} \cap \tilde{Q} = \emptyset$.

2. We next prove $\arg \min_{s \in P \cup Q} d(s, p') \subseteq P, \forall p' \in \tilde{P}$. We do this by contradiction.

If possible, suppose that there exists $p' \in \tilde{P}$ such that $\arg \min_{s \in P \cup Q} d(s, p') \not\subseteq P$. Then there exists $q \in Q$ such that $\min_{s \in P \cup Q} d(s, p') = d(q, p')$. But $d(q, p') \geq \min_{s \in Q} d(s, p') \geq \min_{s \in P \cup Q} d(s, p')$. This implies $\min_{s \in P \cup Q} d(s, p') = \min_{s \in Q} d(s, p')$. Due to the definition of \tilde{Q} in (11) this implies $p' \in \tilde{Q}$. However, we have already shown that $\tilde{P} \cap \tilde{Q} = \emptyset$. This leads to a contradiction. Thus $\arg \min_{s \in P \cup Q} d(s, p') \subseteq P, \forall p' \in \tilde{P}$.

Likewise, we can prove $\arg \min_{s \in P \cup Q} d(s, q') \subseteq Q, \forall q' \in \tilde{Q}$.

3. We next prove $\arg \min_{r' \in \tilde{R}} d(p, r') \subseteq \tilde{P}, \forall p \in P$. We do this by contradiction.

If possible, suppose that there exists $p_3 \in P$ such that $\arg \min_{r' \in \tilde{R}} d(p_3, r') \not\subseteq \tilde{P}$. Then there exists $\rho' \in \tilde{Q}$ such that $\min_{r' \in \tilde{R}} d(p_3, r') = d(p_3, \rho')$.

Again, due to the definition of \tilde{Q} in (11), for any $\rho' \in \tilde{Q}$, there exists $q_3 \in Q$ such that $d(q_3, \rho') = \min_{s \in P \cup Q} d(s, \rho')$.

Thus,

$$\begin{aligned}
 \min_{r' \in \tilde{R}} d(p_3, r') + \min_{s \in P \cup Q} d(s, \rho') &= d(p_3, \rho') + d(q_3, \rho') \\
 &\geq d(p_3, q_3) \quad (\text{triangle inequality}) \\
 &\geq \min_{\substack{p \in P, \\ q \in Q}} d(p, q) \quad (\text{since } p_3 \in P, q_3 \in Q) \\
 \Rightarrow \max_{s \in P \cup Q} \min_{r' \in \tilde{R}} d(s, r') + \max_{r' \in \tilde{R}} \min_{s \in P \cup Q} d(s, r') &\geq \min_{\substack{p \in P, \\ q \in Q}} d(p, q) \\
 \Rightarrow d_H(P \cup Q, \tilde{R}) + \max_{r' \in \tilde{R}} \min_{s \in P \cup Q} d(s, r') &\geq \text{sep}(P, Q).
 \end{aligned}$$

This contradicts assumption (10) of the lemma. Hence there cannot exist $p_3 \in P$ such that $\arg \min_{r' \in \tilde{R}} d(p_3, r') \not\subseteq \tilde{P}$. Thus $\arg \min_{r' \in \tilde{R}} d(p, r') \subseteq \tilde{P}$, $\forall p \in P$.

Likewise, we can prove $\arg \min_{r' \in \tilde{R}} d(q, r') \subseteq \tilde{Q}$, $\forall q \in Q$.

4. Since $\arg \min_{s \in P \cup Q} d(s, p') \subseteq P$, $\forall p' \in \tilde{P}$, we have $\min_{s \in P \cup Q} d(s, p') = \min_{p \in P} d(p, p')$, $\forall p' \in \tilde{P}$. Thus, $\max_{p' \in \tilde{P}} \min_{p \in P} d(p, p') = \max_{p' \in \tilde{P}} \min_{s \in P \cup Q} d(s, p')$.

Likewise, since $\arg \min_{r' \in \tilde{R}} d(p, r') \subseteq \tilde{P}$, $\forall p \in P$, we have $\max_{p \in P} \min_{p' \in \tilde{P}} d(p, p') = \max_{p \in P} \min_{r' \in \tilde{R}} d(p, r')$.

Thus,

$$\begin{aligned}
 d_H(P, \tilde{P}) &= \max \left(\max_{p \in P} \min_{p' \in \tilde{P}} d(p, p'), \max_{p' \in \tilde{P}} \min_{p \in P} d(p, p') \right) \\
 &= \max \left(\max_{p \in P} \min_{r' \in \tilde{R}} d(p, r'), \max_{p' \in \tilde{P}} \min_{s \in P \cup Q} d(s, p') \right) \\
 &\leq \max \left(\max_{s \in P \cup Q} \min_{r' \in \tilde{R}} d(s, r'), \max_{r' \in \tilde{R}} \min_{s \in P \cup Q} d(s, r') \right) \\
 &\quad (\text{since } P \subseteq P \cup Q, \tilde{P} \subseteq \tilde{R}) \\
 &= d_H(P \cup Q, \tilde{R}).
 \end{aligned} \tag{12}$$

Similarly, we can show

$$\begin{aligned}
 d_H(Q, \tilde{Q}) &= \max \left(\max_{q \in Q} \min_{r' \in \tilde{R}} d(q, r'), \max_{q' \in \tilde{Q}} \min_{s \in P \cup Q} d(s, q') \right) \\
 &\leq d_H(P \cup Q, \tilde{R}).
 \end{aligned} \tag{13}$$

Again, from (12) and (13),

$$\begin{aligned}
 \max(d_H(P, \tilde{P}), d_H(Q, \tilde{Q})) &= \max \left(\max_{p \in P} \min_{r' \in \tilde{R}} d(p, r'), \max_{q \in Q} \min_{r' \in \tilde{R}} d(q, r'), \right. \\
 &\quad \left. \max_{p' \in \tilde{P}} \min_{s \in P \cup Q} d(s, p'), \max_{q' \in \tilde{Q}} \min_{s \in P \cup Q} d(s, q') \right) \\
 &= \max \left(\max_{p \in P \cup Q} \min_{r' \in \tilde{R}} d(p, r'), \max_{p' \in \tilde{P} \cup \tilde{Q}} \min_{s \in P \cup Q} d(s, p') \right) \\
 &= d_H(P \cup Q, \tilde{R}) \quad (\text{since } \tilde{P} \cup \tilde{Q} = \tilde{R})
 \end{aligned}$$

5.

$$\begin{aligned}
\text{sep}(\tilde{P}, \tilde{Q}) &\geq \text{sep}(\tilde{P}, Q) - d_H(Q, \tilde{Q}) \quad (\text{using Lemma 4}) \\
&\geq \text{sep}(P, Q) - d_H(P, \tilde{P}) - d_H(Q, \tilde{Q}) \quad (\text{using Lemma 4}) \\
&\geq \text{sep}(P, Q) - 2d_H(P \cup Q, \tilde{R}) \\
&\quad (\text{since } d_H(P, \tilde{P}) \leq d_H(P \cup Q, \tilde{R}) \text{ and } d_H(Q, \tilde{Q}) \leq d_H(P \cup Q, \tilde{R})) \quad \square
\end{aligned}$$

Corollary 1 *If P, Q, \tilde{R} are closed subsets of a metric space (Ψ, d) such that $d_H(P \cup Q, \tilde{R}) < \frac{1}{2} \text{sep}(P, Q)$, then \tilde{R} is a separation-preserving perturbation of P and Q .*

As a consequence, the separation-preserving partition $\{\tilde{P}, \tilde{Q}\}$ of \tilde{R} as defined in (11) satisfies properties '1' to '4' in Lemma 5, as well as property '5' (if (Ψ, d) is a connected path metric space) with an additional inequality:

$$\text{sep}(\tilde{P}, \tilde{Q}) \geq \text{sep}(P, Q) - 2d_H(P \cup Q, \tilde{R}) > 0.$$

Proof The result follows directly from Lemma 5 by observing that

$$\max_{r' \in \tilde{R}} \min_{s \in P \cup Q} d(s, r') + d_H(P \cup Q, \tilde{R}) \leq 2d_H(P \cup Q, \tilde{R}) < \text{sep}(P, Q). \quad \square$$

3 Results on perturbation upper bounds

Throughout this section we use the notations and conventions described in Definition 1.

3.1 Elementary results on spectrum perturbation

In this section we provide some elementary results relating the norm of the matrix perturbation and the perturbation of eigenvalues and eigenvectors.

Lemma 6 *Define $D \in \mathbb{C}^{n \times n}$ such that $D_{jj'} = (\tilde{\lambda}_{j'} - \lambda_j) \mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}$. Then*

$$D = U^\dagger (\tilde{M} - M) \tilde{U}. \quad (14)$$

Equivalently,

$$(\tilde{\lambda}_{j'} - \lambda_j) \mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'} = \mathbf{u}_j^\dagger (\tilde{M} - M) \tilde{\mathbf{u}}_{j'}, \quad \forall j, j' \in N. \quad (15)$$

The latter relation in fact holds even when \tilde{M} is not normal but $\tilde{\mathbf{u}}_{j'}$ is simply a right eigenvector of \tilde{M} with the corresponding eigenvalue $\tilde{\lambda}_{j'}$.

Proof First we note that since M is normal with \mathbf{u}_j , a right eigenvector and the corresponding eigenvalue λ_j , \mathbf{u}_j^\dagger is a left eigenvector of M with the same eigenvalue. Thus,

$$\begin{aligned}
\mathbf{u}_j^\dagger (\tilde{M} - M) \tilde{\mathbf{u}}_{j'} &= \mathbf{u}_j^\dagger \tilde{M} \tilde{\mathbf{u}}_{j'} - \mathbf{u}_j^\dagger M \tilde{\mathbf{u}}_{j'} \\
&= \mathbf{u}_j^\dagger \tilde{\lambda}_{j'} \tilde{\mathbf{u}}_{j'} - \lambda_j \mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'} \\
&= (\tilde{\lambda}_{j'} - \lambda_j) \mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}.
\end{aligned}$$

This proves (15).

We note that if both M and \tilde{M} are normal, the L.H.S. of (15) is the (j, j') th element of $U^\dagger(\tilde{M} - M)\tilde{U}$ and the R.H.S. is $D_{jj'}$. \square

Corollary 2

$$\begin{aligned}\|\tilde{M} - M\|_2^2 &\geq \|(\tilde{M} - M)\tilde{\mathbf{u}}_{j'}\|_2^2 = \sum_{j=1}^n |\tilde{\lambda}_{j'} - \lambda_j|^2 |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2, \quad \forall j' \in N \\ \|\tilde{M} - M\|_2^2 &\geq \|(\tilde{M} - M)\mathbf{u}_j\|_2^2 = \sum_{j'=1}^n |\tilde{\lambda}_{j'} - \lambda_j|^2 |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2, \quad \forall j \in N.\end{aligned}\tag{16}$$

The first relation holds even when \tilde{M} is not normal, while the second relation holds even when M is not normal.

Proof The inequalities follow from the definition of induced 2-norm for matrices.

When M is normal, $\{\mathbf{u}_j\}_{j \in N}$ forms an orthonormal basis in \mathbb{C}^n . Noting that (15) is a scalar equation, multiplying on both sides with \mathbf{u}_j and summing over j , we get

$$\begin{aligned}\sum_{j=1}^n (\tilde{\lambda}_{j'} - \lambda_j) \mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'} \mathbf{u}_j &= \sum_{j=1}^n \mathbf{u}_j (\mathbf{u}_j^\dagger (\tilde{M} - M) \tilde{\mathbf{u}}_{j'}) \\ &= \left(\sum_{j=1}^n \mathbf{u}_j \mathbf{u}_j^\dagger \right) (\tilde{M} - M) \tilde{\mathbf{u}}_{j'} \\ &= I(\tilde{M} - M) \tilde{\mathbf{u}}_{j'}.\end{aligned}$$

Taking the 2-norm on both sides of the above gives the first equality.

Switching the roles of tilde and nontilde terms in Lemma 6 and the above gives the second relation. \square

Corollary 3

1

$$\begin{aligned}\|\tilde{M} - M\|_2 &\geq \|(\tilde{M} - M)\tilde{\mathbf{u}}_{j'}\|_2 \geq \min_{j \in N} |\tilde{\lambda}_{j'} - \lambda_j|, \quad \forall j' \in N, \quad \text{and} \\ \|\tilde{M} - M\|_2 &\geq \|(\tilde{M} - M)\mathbf{u}_j\|_2 \geq \min_{j' \in N} |\tilde{\lambda}_{j'} - \lambda_j|, \quad \forall j \in N.\end{aligned}$$

The first relation holds even when \tilde{M} is not normal, while the second relation holds even when M is not normal.

2 The following results are a consequence of the Bauer–Fike theorem for normal matrices [9]:

$$\begin{aligned}\|\tilde{M} - M\|_2 &\geq \max_{j \in N} \|(\tilde{M} - M)\tilde{\mathbf{u}}_{j'}\|_2 \geq \max_{j' \in N} \min_{j \in N} |\tilde{\lambda}_{j'} - \lambda_j|, \\ \|\tilde{M} - M\|_2 &\geq \max_{j \in N} \|(\tilde{M} - M)\mathbf{u}_j\|_2 \geq \max_{j \in N} \min_{j' \in N} |\tilde{\lambda}_{j'} - \lambda_j|.\end{aligned}$$

Once again, the first relation holds even when \tilde{M} is not normal, while the second relation holds even when M is not normal.

Proof From the result of Corollary 2, when M is normal (and \tilde{M} is not necessarily normal), for all $j' \in N$,

$$\begin{aligned}\|\tilde{M} - M\|_2^2 &\geq \|(\tilde{M} - M)\tilde{\mathbf{u}}_{j'}\|_2^2 \\ &= \sum_{j=1}^n |\tilde{\lambda}_{j'} - \lambda_j|^2 |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2 \\ &\geq \min_{j \in N} |\tilde{\lambda}_{j'} - \lambda_j|^2 \sum_{j=1}^n |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2 \\ &= \min_{j \in N} |\tilde{\lambda}_{j'} - \lambda_j|^2 \|\tilde{\mathbf{u}}_{j'}\|^2 \quad (\text{since } \{\mathbf{u}_j\}_{j \in N} \text{ forms an orthonormal basis}) \\ &= \min_{j \in N} |\tilde{\lambda}_{j'} - \lambda_j|^2.\end{aligned}$$

Since this is true for any $j' \in N$, it follows that $\|\tilde{M} - M\|_2 \geq \max_{j' \in N} \min_{j \in N} |\tilde{\lambda}_{j'} - \lambda_j|$.

A similar set of the results can be derived with the tilde and nontilde terms exchanged. \square

3.2 Distance between invariant subspaces of normal matrices with partitioned spectra

Suppose $J, \tilde{J} \subseteq N$ such that $|J| = |\tilde{J}| = q$. We are interested in understanding how much the invariant space $\text{span}(\mathbf{u}_J)$ of M differs from the invariant space $\text{span}(\tilde{\mathbf{u}}_{\tilde{J}})$ of \tilde{M} . The results in this section are variations and modest improvements on the Davis–Kahan $\sin \Theta$ theorem [1] (see Section VIII.3 of [2] for example). In Proposition 1 and the two corollaries that follow, we present results of the form

$$d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) \leq \mathcal{F}(\tilde{M} - M, \mathbf{u}_N, \lambda_N, \tilde{\lambda}_N; J, \tilde{J}),$$

where \mathcal{F} is a function specific to the exact statement of the proposition or corollary.

For a given invariant subspace $\text{span}(\mathbf{u}_J)$ of M , we can consider all the possible q -dimensional invariant subspaces of \tilde{J} and choose the one that is closest to $\text{span}(\mathbf{u}_J)$ as its perturbation. As a consequence, for any of these results, we can write

$$\min_{\tilde{J} \in S_{q,n}} d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) \leq \min_{\tilde{J} \in S_{q,n}} \mathcal{F}(\tilde{M} - M, \mathbf{u}_N, \lambda_N, \tilde{\lambda}_N; J, \tilde{J}),$$

where $S_{q,n}$ is the set of all q -element subsets of $N = \{1, 2, \dots, n\}$. This gives a combinatorial means of finding the q -dimensional invariant subspace of \tilde{M} that is closest to $\text{span}(\mathbf{u}_J)$.

Definition 4 For $a, b, c \in \mathbb{R}$ with $a \leq \min(b, c)$, we define

$$[a, \min(b, c_-)] = \begin{cases} [a, b] & \text{if } c > b, \\ [a, c] & \text{if } c \leq b. \end{cases}$$

Proposition 1 For any $J, \tilde{J} \subseteq N$ with $|J| = |\tilde{J}| = q$,

$$d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) \leq \sqrt{\frac{1}{q} \sum_{j \in J} \frac{\|(\tilde{M} - M)\mathbf{u}_j\|_2^2 - \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}} \quad (17)$$

for any $\kappa_j \in [0, \min(1, (\frac{\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2})_-)]$, $j \in J$.

The tightest bound in (17) is obtained by choosing

$$\kappa_j = \begin{cases} 0 & \text{if } \|(\tilde{M} - M)\mathbf{u}_j\|_2 \geq \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|, \\ 1 & \text{if } \|(\tilde{M} - M)\mathbf{u}_j\|_2 < \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|. \end{cases} \quad (18)$$

Proof From Corollary 2, for all $j \in N$,

$$\begin{aligned} & \|(\tilde{M} - M)\mathbf{u}_j\|_2^2 \\ &= \sum_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2 |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2 + \sum_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2 \\ &\geq \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2 \sum_{j' \in \tilde{J}} |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2 + \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 \sum_{j' \in \tilde{J}^c} |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2 \\ &= \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2 \left(1 - \sum_{j' \in \tilde{J}^c} |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2\right) + \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 \sum_{j' \in \tilde{J}^c} |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2 \\ &\geq \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2 \left(1 - \sum_{j' \in \tilde{J}^c} |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2\right) + \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 \sum_{j' \in \tilde{J}^c} |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2 \end{aligned} \quad (19)$$

for any $\kappa_j \in [0, 1]$.

$$\begin{aligned} &\Rightarrow \left(\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2 \right) \sum_{j' \in \tilde{J}^c} |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2 \\ &\leq \|(\tilde{M} - M)\mathbf{u}_j\|_2^2 - \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2 \end{aligned} \quad (20)$$

$$\begin{aligned} &\Rightarrow \sum_{j' \in \tilde{J}^c} |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2 \leq \frac{\|(\tilde{M} - M)\mathbf{u}_j\|_2^2 - \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2} \\ &\text{for any } \kappa_j \in [0, \min(1, (\frac{\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2})_-)]. \end{aligned} \quad (21)$$

In the last step, we ensured that $\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2$ is positive by restricting the domain of κ_j appropriately.

Thus, from (21) we have

$$\begin{aligned} & (d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})))^2 \\ &= \frac{1}{q} \sum_{\substack{j \in J \\ j' \in \tilde{J}^c}} |\mathbf{u}_j^\dagger \tilde{\mathbf{u}}_{j'}|^2 \quad (\text{due to Lemma 1.2}) \end{aligned}$$

$$\leq \frac{1}{q} \sum_{j \in J} \frac{\|(\tilde{M} - M)\mathbf{u}_j\|_2^2 - \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2} \quad (22)$$

for any $\kappa_j \in [0, \min(1, (\frac{\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2})_-)]$, $j \in J$.

Additionally, we note that

$$\begin{aligned} \|(\tilde{M} - M)\mathbf{u}_j\|_2 < \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j| &\Rightarrow \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j| > \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j| \\ &\left(\text{since, due to Corollary 3, } \|(\tilde{M} - M)\mathbf{u}_j\|_2 \geq \min_{j' \in N} |\tilde{\lambda}_{j'} - \lambda_j| \right). \end{aligned}$$

Thus, when $\|(\tilde{M} - M)\mathbf{u}_j\|_2 < \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|$, the valid domain of κ_j is $[0, 1]$. The statement about the tightest bound then follows from the fact that the function $f(\kappa) = \frac{a - \kappa c}{b - \kappa c}$, $\kappa \in [0, d]$ (with $d < \frac{b}{c}$) is minimized with $\kappa = 0$ when $a \geq b$, and with $\kappa = d$ when $a < b$. \square

The key achievement in the above proposition is to provide an upper bound on the distance (in terms of d_{sp}) between the invariant subspaces $\text{span}(\mathbf{u}_J)$ and $\text{span}(\tilde{\mathbf{u}}_{\tilde{J}})$ in terms of the distance between the matrices M and \tilde{M} and their eigenvalues. For a given/fixed matrix perturbation $(\tilde{M} - M)$ and appropriately chosen \tilde{J} , inequality (17) can be interpreted as a relation between the perturbation in the eigenvalues $\{\lambda_j | j \in J\}$ and the perturbation in the invariant space $\text{span}(\mathbf{u}_J)$. This relationship, in general, can be expected to be an inverse one – with higher perturbation in the eigenvalues we will have a lower (upper bound on the) perturbation in the invariant space, and vice versa.

It is easy to note that the equality in (17) holds when

- (i) $\frac{\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2} > 1$, $\forall j \in J$, allowing us to choose $\kappa_j = 1$, $\forall j \in J$, and
- (ii)

$$\begin{aligned} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_{j_1}| &= \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_{j_2}|, \quad \forall j_1, j_2 \in J, \\ \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_{j_1}| &= \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_{j_2}|, \quad \forall j_1, j_2 \in J^c. \end{aligned}$$

(These conditions hold, for example, when $\tilde{\lambda}_{\tilde{J}}$ and $\tilde{\lambda}_{\tilde{J}^c}$ are small translations of λ_J and λ_{J^c} respectively in \mathbb{C} .)

In Proposition 1, without loss of generality, we can interchange the roles of J and J^c (likewise \tilde{J} and \tilde{J}^c). Observing that $\text{span}(\mathbf{u}_{J^c})$ and $\text{span}(\tilde{\mathbf{u}}_{\tilde{J}^c})$ are $(n - q)$ dimensional subspaces of \mathbb{C}^n which are orthogonal complements of $\text{span}(\mathbf{u}_J)$ and $\text{span}(\tilde{\mathbf{u}}_{\tilde{J}})$ respectively, we then obtain

$$\begin{aligned} d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) &= \sqrt{\frac{n - q}{q}} d_{\text{sp}}(\text{span}(\mathbf{u}_{J^c}), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}^c})) \quad (\text{due to Lemma 2.3}) \\ &\leq \sqrt{\frac{1}{q} \sum_{j \in J^c} \frac{\|(\tilde{M} - M)\mathbf{u}_j\|_2^2 - \kappa_j \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_j \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2}} \quad (23) \end{aligned}$$

for any $\kappa_j \in [0, \min(1, (\frac{\min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2})_-)]$, $j \in J^c$.

Corollary 4 For any $\kappa_J \in [0, \min(1, (\frac{\min_{j \in J, j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\max_{j \in J} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2})_-)]$ and

$$\kappa_{J^c} \in [0, \min(1, (\frac{\min_{j \in J^c, j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\max_{j \in J^c} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2})_-)],$$

1.

$$\begin{aligned} & d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) \\ & \leq \frac{1}{q} \min \left(\sqrt{\frac{\sum_{j \in J} \|(\tilde{M} - M)\mathbf{u}_j\|_2^2 - \kappa_J \sum_{j \in J} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j \in J} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_J \max_{j \in J} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}}, \right. \\ & \quad \left. \sqrt{\frac{\sum_{j \in J^c} \|(\tilde{M} - M)\mathbf{u}_j\|_2^2 - \kappa_{J^c} \sum_{j \in J^c} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j \in J^c} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_{J^c} \max_{j \in J^c} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2}} \right). \end{aligned} \quad (24)$$

2.

$$\begin{aligned} & d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) \\ & \leq \sqrt{\frac{\frac{1}{q} \left(\|\tilde{M} - M\|_F^2 - \left(\kappa_J \sum_{j \in J} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2 + \kappa_{J^c} \sum_{j \in J^c} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 \right) \right)}{\min_{j \in J} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \left(\kappa_J \max_{j \in J} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2 + \kappa_{J^c} \max_{j \in J^c} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 \right)}}} \end{aligned} \quad (25)$$

Proof With $\kappa_j \in [0, \min(1, (\frac{\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2})_-)]$, $j \in J$,

$$\begin{aligned} & q(d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})))^2 \\ & \leq \sum_{j \in J} \frac{\|(\tilde{M} - M)\mathbf{u}_j\|_2^2 - \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2} \text{ (due to Proposition 1)} \\ & \leq \frac{\sum_{j \in J} \|(\tilde{M} - M)\mathbf{u}_j\|_2^2 - \sum_{j \in J} \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j \in J} (\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2)} \\ & \quad \left(\text{since } \sum_{k \in S} \frac{c_k}{d_k} \leq \frac{\sum_{k \in S} c_k}{\min_{k \in S} d_k} \right) \\ & \leq \frac{\sum_{j \in J} \|(\tilde{M} - M)\mathbf{u}_j\|_2^2 - \sum_{j \in J} \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j \in J} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \max_{j \in J} \kappa_j \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2} \\ & \quad \left(\min_{k \in S} (c_k - d_k) \geq \min_{k \in S} c_k - \max_{k \in S} d_k \right). \end{aligned} \quad (26)$$

We next choose $\kappa_j = \kappa_k$, $\forall j, k \in J$ and denote this value by

$$\begin{aligned} & \kappa_J \in \bigcap_{j \in J} \left[0, \min \left(1, \left(\frac{\min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2} \right)_- \right) \right] \\ & \supseteq \left[0, \min \left(1, \left(\frac{\min_{j \in J} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\max_{j \in J} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2} \right)_- \right) \right]. \end{aligned}$$

Thus,

$$\begin{aligned} & q(d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})))^2 \\ & \leq \frac{\sum_{j \in J} \|(\tilde{M} - M)\mathbf{u}_j\|_2^2 - \kappa_J \sum_{j \in J} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{\substack{j \in J \\ j' \in \tilde{J}^c}} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_J \max_{j \in J} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2} \end{aligned} \quad (27)$$

for any $\kappa_J \in [0, \min(1, (\frac{\min_{j \in J, j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\max_{j \in J} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}) -)]$.

By interchanging the roles of J and J^c (accordingly, \tilde{J} and \tilde{J}^c) and noting that $\text{span}(\mathbf{u}_{J^c})$ and $\text{span}(\tilde{\mathbf{u}}_{\tilde{J}^c})$ are $(n - q)$ dimensional sub-spaces of \mathbb{C}^n , we get

$$\begin{aligned} & (n - q)(d_{\text{sp}}(\text{span}(\mathbf{u}_{J^c}), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}^c})))^2 \\ & \leq \frac{\sum_{j \in J^c} \|(\tilde{M} - M)\mathbf{u}_j\|_2^2 - \kappa_{J^c} \sum_{j \in J^c} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{\substack{j \in J^c \\ j' \in \tilde{J}}} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_{J^c} \max_{j \in J^c} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2} \end{aligned} \quad (28)$$

for any $\kappa_{J^c} \in [0, \min(1, (\frac{\min_{j \in J^c, j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\max_{j \in J^c} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}) -)]$.

On the other hand, since $\text{span}(\mathbf{u}_J)$ and $\text{span}(\mathbf{u}_{J^c})$ are orthogonal complements (likewise, $\text{span}(\mathbf{u}_{\tilde{J}})$ and $\text{span}(\mathbf{u}_{\tilde{J}^c})$ are orthogonal complements), using Lemma 2, we can write (28) as

$$\begin{aligned} & q(d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})))^2 \\ & \leq \frac{\sum_{j \in J^c} \|(\tilde{M} - M)\mathbf{u}_j\|_2^2 - \kappa_{J^c} \sum_{j \in J^c} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\min_{\substack{j \in J^c \\ j' \in \tilde{J}}} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_{J^c} \max_{j \in J^c} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2} \end{aligned} \quad (29)$$

for any $\kappa_{J^c} \in [0, \min(1, (\frac{\min_{j \in J^c, j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}{\max_{j \in J^c} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2}) -)]$.

Combining (27) and (29) proves part ‘1’.

Again, adding (27) and (29), we have

$$\begin{aligned} & q\left(\min_{\substack{j \in J \\ j' \in \tilde{J}^c}} |\tilde{\lambda}_{j'} - \lambda_j|^2 + \min_{\substack{j \in J^c \\ j' \in \tilde{J}}} |\tilde{\lambda}_{j'} - \lambda_j|^2\right. \\ & \quad \left. - \kappa_J \max_{j \in J} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_{J^c} \max_{j \in J^c} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2\right) \\ & \quad \times (d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})))^2 \\ & \leq \sum_{j \in J} \|(\tilde{M} - M)\mathbf{u}_j\|_2^2 + \sum_{j \in J^c} \|(\tilde{M} - M)\mathbf{u}_j\|_2^2 \\ & \quad - \kappa_J \sum_{j \in J} \min_{j' \in \tilde{J}} |\tilde{\lambda}_{j'} - \lambda_j|^2 - \kappa_{J^c} \sum_{j \in J^c} \min_{j' \in \tilde{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2. \end{aligned}$$

The part ‘2’ of the result then follows by observing that

$$\sum_{j \in J} \|(\tilde{M} - M)\mathbf{u}_j\|_2^2 + \sum_{j \in J^c} \|(\tilde{M} - M)\mathbf{u}_j\|_2^2 = \|(\tilde{M} - M)U\|_F^2 = \|\tilde{M} - M\|_F^2.$$

□

Corollary 5 (Generalized Davis–Kahan [1] $\sin \Theta$ theorem for normal matrices – see Section VIII.3 of [2])

1.

$$d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) \leq \frac{\min(1, \sqrt{\frac{n-q}{q}})}{\max(\text{sep}(\lambda_J, \tilde{\lambda}_{\tilde{J}^c}), \text{sep}(\lambda_{J^c}, \tilde{\lambda}_{\tilde{J}}))} \|\tilde{M} - M\|_2;$$

2.

$$\begin{aligned} d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) &\leq \frac{\frac{1}{\sqrt{q}} \|\tilde{M} - M\|_F}{\sqrt{\text{sep}(\lambda_J, \tilde{\lambda}_{\tilde{J}^c})^2 + \text{sep}(\lambda_{J^c}, \tilde{\lambda}_{\tilde{J}})^2}} \\ &\leq \sqrt{\frac{n/q}{\text{sep}(\lambda_J, \tilde{\lambda}_{\tilde{J}^c})^2 + \text{sep}(\lambda_{J^c}, \tilde{\lambda}_{\tilde{J}})^2}} \|\tilde{M} - M\|_2. \end{aligned}$$

Proof In (27), setting $\kappa_J = 0$, we get

$$(d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})))^2 \leq \frac{\frac{1}{q} \sum_{j \in J} \|(\tilde{M} - M)\mathbf{u}_j\|_2^2}{\text{sep}(\lambda_J, \tilde{\lambda}_{\tilde{J}^c})^2} \leq \frac{\|(\tilde{M} - M)\|_2^2}{\text{sep}(\lambda_J, \tilde{\lambda}_{\tilde{J}^c})^2}.$$

Interchanging the roles of the tilde and nontilde terms in this result, we analogously obtain

$$(d_{\text{sp}}(\text{span}(\tilde{\mathbf{u}}_{\tilde{J}}, \text{span}(\mathbf{u}_J)))^2 \leq \frac{\|(\tilde{M} - M)\|_2^2}{\text{sep}(\tilde{\lambda}_{\tilde{J}}, \lambda_{J^c})^2}.$$

The above two together give

$$(d_{\text{sp}}(\text{span}(\tilde{\mathbf{u}}_{\tilde{J}}, \text{span}(\mathbf{u}_J)))^2 \leq \frac{\|(\tilde{M} - M)\|_2^2}{\max(\text{sep}(\lambda_J, \tilde{\lambda}_{\tilde{J}^c}), \text{sep}(\lambda_{J^c}, \tilde{\lambda}_{\tilde{J}}))^2}. \quad (30)$$

In the above inequality, interchanging the roles of J and J^c (accordingly, \tilde{J} and \tilde{J}^c) and observing that by Lemma 2 $d_{\text{sp}}(\text{span}(\tilde{\mathbf{u}}_{\tilde{J}}, \text{span}(\mathbf{u}_J)) = \sqrt{\frac{n-q}{q}} d_{\text{sp}}(\text{span}(\tilde{\mathbf{u}}_{\tilde{J}^c}, \text{span}(\mathbf{u}_{J^c}))$, we obtain

$$(d_{\text{sp}}(\text{span}(\tilde{\mathbf{u}}_{\tilde{J}}, \text{span}(\mathbf{u}_J)))^2 \leq \frac{n-q}{q} \frac{\|(\tilde{M} - M)\|_2^2}{\max(\text{sep}(\lambda_J, \tilde{\lambda}_{\tilde{J}^c}), \text{sep}(\lambda_{J^c}, \tilde{\lambda}_{\tilde{J}}))^2}. \quad (31)$$

(30) and (31) together conclude the proof of part ‘1’.

The second result follows directly from part ‘2’ of Corollary 4 by setting $\kappa_J = \kappa_{J^c} = 0$ and using the fact that for $Q \in \mathbb{C}^{n \times n}$, $\|Q\|_F \leq \sqrt{n} \|Q\|_2$. \square

3.3 Bound on perturbation of invariant subspace of a normal matrix with well-clustered spectrum

In this section we specialize the earlier results for the situation when λ_J and λ_{J^c} are well-clustered (i.e., the separation between them is large) compared to the perturbation $(\tilde{M} - M)$. In the following lemma we outline the conditions under which the perturbed eigenvalues $\tilde{\lambda}_N$ will also remain well clustered.

Lemma 7 For any $J \subseteq N$, define $J^c = N - J$. If $\|\tilde{M} - M\|_2 < \frac{1}{2} \text{sep}(\lambda_J, \lambda_{J^c})$, then:

1. $\tilde{\lambda}_N$ is a separation-preserving perturbation of λ_J and λ_{J^c} . More explicitly, defining

$$\begin{aligned}\widehat{J} &= \left\{ j' \mid \min_{j \in N} |\tilde{\lambda}_{j'} - \lambda_j| = \min_{j \in J} |\tilde{\lambda}_{j'} - \lambda_j| \right\} \quad \text{and} \\ \widehat{J}^c &= \left\{ j' \mid \min_{j \in N} |\tilde{\lambda}_{j'} - \lambda_j| = \min_{j \in J^c} |\tilde{\lambda}_{j'} - \lambda_j| \right\}\end{aligned}\quad (32)$$

makes $\{\tilde{\lambda}_{\widehat{J}}, \tilde{\lambda}_{\widehat{J}^c}\}$ a separation-preserving partition of $\tilde{\lambda}_N$, with

$$\text{sep}(\tilde{\lambda}_{\widehat{J}}, \tilde{\lambda}_{\widehat{J}^c}) > \text{sep}(\lambda_J, \lambda_{J^c}) - 2\|\tilde{M} - M\|_2.$$

2. $|\tilde{\lambda}_{\widehat{J}}| = |\lambda_J|$ (equivalently, $|\tilde{\lambda}_{\widehat{J}^c}| = |\lambda_{J^c}|$), where $|\cdot|$ denotes the number of elements in the multi-sets (recall that λ_J and $\tilde{\lambda}_{\widehat{J}}$ are multi-sets, allowing them to contain multiple copies of nondistinct eigenvalues, if any, of M and \tilde{M} respectively).

Proof

1. We first observe that

$$\|\tilde{M} - M\|_2 \geq \max \left(\max_{j \in N} \min_{j' \in N} |\tilde{\lambda}_{j'} - \lambda_j|, \max_{j' \in N} \min_{j \in N} |\tilde{\lambda}_{j'} - \lambda_j| \right) = d_H(\lambda_N, \tilde{\lambda}_N). \quad (33)$$

As a consequence, $d_H(\lambda_N, \tilde{\lambda}_N) \leq \|\tilde{M} - M\|_2 < \frac{1}{2} \text{sep}(\lambda_J, \lambda_{J^c})$. Then the proof of the first part follows directly from Corollary 1 by setting $P = \lambda_J$, $Q = \lambda_{J^c}$ and $\tilde{R} = \tilde{\lambda}_N$.

2. We prove the second part by contradiction.

If possible, let $|\tilde{\lambda}_{\widehat{J}}| \neq |\lambda_J|$. Without loss of generality, we will assume $|\tilde{\lambda}_{\widehat{J}}| < |\lambda_J|$ (if the $|\tilde{\lambda}_{\widehat{J}}| > |\lambda_J|$, we can show the contradiction for $|\tilde{\lambda}_{\widehat{J}^c}| < |\lambda_{J^c}|$ instead).

Define a path $\overline{M} : [0, 1] \rightarrow \mathbb{R}^{n \times n}$ connecting M and \tilde{M} as

$$\overline{M}(t) = t\tilde{M} + (1-t)M.$$

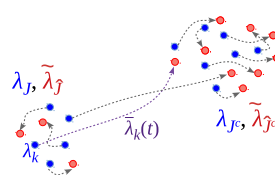
Although $\overline{M}(t)$ is not necessarily normal for all t , its characteristic equation is a degree- n polynomial equation in its eigenvalue with coefficient of the highest degree term equal to 1 and other coefficients being polynomials in t . Since the roots of such a polynomial are continuous functions of the coefficients, the eigenvalues of $\overline{M}(t)$ are continuous functions of t . Thus, we define $\overline{\lambda}_j : [0, 1] \rightarrow \mathbb{C}$ to be the paths of the eigenvalues such that $\overline{\lambda}_j(0) = \lambda_j$ for all $j \in \{1, 2, \dots, n\}$. $\overline{\lambda}_j(1)$ are the eigenvalues of $\overline{M}(1) = \tilde{M}$, so that $\overline{\lambda}_j(1) = \tilde{\lambda}_{\sigma(j)}$ for some permutation $\sigma : \{1, 2, \dots, n\} \rightarrow \{1, 2, \dots, n\}$ (see Fig. 4).

Since $|\tilde{\lambda}_{\widehat{J}}| < |\lambda_J|$, there exists at least one $k \in J$ (with $\lambda_k = \overline{\lambda}_k(0) \in \lambda_J$) such that $\overline{\lambda}_k(1) \notin \tilde{\lambda}_{\widehat{J}}$ (equivalently, $\overline{\lambda}_k(1) \in \tilde{\lambda}_{\widehat{J}^c}$).

Define $g(t) = \min_{j \in J} |\overline{\lambda}_k(t) - \lambda_j|$ and $h(t) = \min_{j \in J^c} |\overline{\lambda}_k(t) - \lambda_j|$. Thus,

$$g(0) = \min_{j \in J} |\overline{\lambda}_k(0) - \lambda_j| = \min_{j \in J} |\lambda_k - \lambda_j| = 0 \quad (\text{since } \lambda_k \in \lambda_J) \leq h(0).$$

Figure 4 Illustration for the proof of Lemma 7



Again,

$$\begin{aligned}
 h(1) &= \min_{j \in J^c} |\bar{\lambda}_k(1) - \lambda_j| \\
 &\leq \min_{j \in J} |\bar{\lambda}_k(1) - \lambda_j| \quad \left(\text{since } \bar{\lambda}_k(1) \in \tilde{\lambda}_{J^c}, \text{ from the definition of } \widehat{J}^c, \right. \\
 &\quad \left. \min_{j \in J^c} |\bar{\lambda}_k(1) - \lambda_j| = \min_{j \in N} |\bar{\lambda}_k(1) - \lambda_j| \right) \\
 &= g(1).
 \end{aligned}$$

Thus, by intermediate value theorem, there exists $t' \in [0, 1]$ such that $g(t') = h(t')$.

That is, $\min_{j \in J} |\bar{\lambda}_k(t') - \lambda_j| = \min_{j \in J^c} |\bar{\lambda}_k(t') - \lambda_j|$. Equivalently,

$$\text{sep}(\lambda_J, \{\bar{\lambda}_k(t')\}) = \text{sep}(\lambda_{J^c}, \{\bar{\lambda}_k(t')\}) \quad \text{for some } t' \in [0, 1]. \quad (34)$$

Now,

$$\begin{aligned}
 \|\bar{M}(t') - M\|_2 &\geq \min_{j \in N} |\bar{\lambda}_k(t') - \lambda_j| \quad (\text{Corollary 3.1}) \\
 &= \min \left(\min_{j \in J} |\bar{\lambda}_k(t') - \lambda_j|, \min_{j \in J^c} |\bar{\lambda}_k(t') - \lambda_j| \right) \\
 &= \frac{1}{2} \left(\min_{j \in J} |\bar{\lambda}_k(t') - \lambda_j| + \min_{j \in J^c} |\bar{\lambda}_k(t') - \lambda_j| \right) \\
 &\quad \left(\text{since from (34), } \min_{j \in J} |\bar{\lambda}_k(t') - \lambda_j| = \min_{j \in J^c} |\bar{\lambda}_k(t') - \lambda_j| \right) \\
 &= \frac{1}{2} \left(\text{sep}(\lambda_J, \{\bar{\lambda}_k(t')\}) + \text{sep}(\lambda_{J^c}, \{\bar{\lambda}_k(t')\}) + \text{diam}(\{\bar{\lambda}_k(t')\}) \right) \\
 &\quad (\text{since the diameter of a point is zero}) \\
 &\geq \frac{1}{2} \text{sep}(\lambda_J, \lambda_{J^c}) \quad (\text{using Lemma 3}).
 \end{aligned} \quad (35)$$

However, $\|\bar{M}(t') - M\|_2 = t' \|\tilde{M} - M\|_2 < t' \frac{1}{2} \text{sep}(\lambda_J, \lambda_{J^c}) \leq \frac{1}{2} \text{sep}(\lambda_J, \lambda_{J^c})$. We thus end up with a contradiction. \square

In the following propositions, we express the upper bounds on $d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\widehat{J}}))$ in terms of $(\tilde{M} - M)$ and nontilde terms only.

Proposition 2 For any $J \subseteq N$ such that $|J| = q$, define $J^c = N - J$. If $\|\tilde{M} - M\|_2 < \frac{1}{2} \text{sep}(\lambda_J, \lambda_{J^c})$,

1.

$$\begin{aligned}
 &d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\widehat{J}})) \\
 &\leq \frac{1}{\sqrt{q}} \min \left(\sqrt{\sum_{j \in J} \left(\frac{\|(\tilde{M} - M)\mathbf{u}_j\|_2}{\min_{k \in J^c} |\lambda_k - \lambda_j| - \|\tilde{M} - M\|_2} \right)^2}, \right. \\
 &\quad \left. \sqrt{\sum_{j \in J^c} \left(\frac{\|(\tilde{M} - M)\mathbf{u}_j\|_2}{\min_{k \in J} |\lambda_k - \lambda_j| - \|\tilde{M} - M\|_2} \right)^2} \right)
 \end{aligned} \quad (36)$$

$$\leq \min\left(1, \sqrt{\frac{n-q}{q}}\right) \frac{\|\tilde{M} - M\|_2}{\text{sep}(\lambda_J, \lambda_{J^c}) - \|\tilde{M} - M\|_2}. \quad (37)$$

2.

$$d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\hat{J}})) \leq \frac{\frac{1}{\sqrt{2q}} \|\tilde{M} - M\|_F}{\text{sep}(\lambda_J, \lambda_{J^c}) - \|\tilde{M} - M\|_2}, \quad (38)$$

where \hat{J} and \hat{J}^c are as defined in (32).

Proof For any $j \in J$,

$$\begin{aligned} \min_{j' \in \hat{J}^c} |\tilde{\lambda}_{j'} - \lambda_j| &= \text{sep}(\{\lambda_j\}, \tilde{\lambda}_{\hat{J}^c}) \\ &\geq \text{sep}(\{\lambda_j\}, \lambda_{J^c}) - d_H(\lambda_{J^c}, \tilde{\lambda}_{\hat{J}^c}) \quad (\text{due to Lemma 4}) \\ &\geq \text{sep}(\{\lambda_j\}, \lambda_{J^c}) - d_H(\lambda_N, \tilde{\lambda}_N) \\ &\quad (\text{due to Lemma 5.4, } d_H(\lambda_{J^c}, \tilde{\lambda}_{\hat{J}^c}) \leq d_H(\lambda_N, \tilde{\lambda}_N)) \\ &\geq \text{sep}(\{\lambda_j\}, \lambda_{J^c}) - \|\tilde{M} - M\|_2 \quad (\text{using (33)}) \\ &= \min_{k \in J^c} |\lambda_k - \lambda_j| - \|\tilde{M} - M\|_2. \end{aligned} \quad (39)$$

Thus, in Proposition 1 choosing $\kappa_j = 0, \forall j \in J$, we get

$$\begin{aligned} (d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\hat{J}})))^2 &\leq \frac{1}{q} \sum_{j \in J} \frac{\|(\tilde{M} - M)\mathbf{u}_j\|_2^2}{\min_{j' \in \hat{J}^c} |\tilde{\lambda}_{j'} - \lambda_j|^2} \\ &\leq \frac{1}{q} \sum_{j \in J} \frac{\|(\tilde{M} - M)\mathbf{u}_j\|_2^2}{(\min_{k \in J^c} |\lambda_k - \lambda_j| - \|\tilde{M} - M\|_2)^2} \\ &\leq \frac{\frac{1}{q} \sum_{j \in J} \|(\tilde{M} - M)\mathbf{u}_j\|_2^2}{\min_{j \in J} (\min_{k \in J^c} |\lambda_k - \lambda_j| - \|\tilde{M} - M\|_2)^2} \\ &\quad \left(\text{since } \sum_{k \in S} \frac{c_k}{d_k} \leq \frac{\sum_{k \in S} c_k}{\min_{k \in S} d_k} \right) \\ &\leq \frac{\|\tilde{M} - M\|_2^2}{(\text{sep}(\lambda_J, \lambda_{J^c}) - \|\tilde{M} - M\|_2)^2} \\ &\quad \left(\text{since } \|\tilde{M} - M\|_2 \geq \|(\tilde{M} - M)\mathbf{u}_j\|_2 \text{ and} \right. \\ &\quad \left. \min_{k \in J} (c_k - \alpha)^2 = \left(\min_{k \in J} c_k - \alpha \right)^2 \right). \end{aligned} \quad (40)$$

In the above, switching the roles of J and J^c (likewise, \hat{J} and \hat{J}^c) and noting that $\text{span}(\mathbf{u}_{J^c})$ and $\text{span}(\tilde{\mathbf{u}}_{\hat{J}^c})$ are $(n - q)$ -dimensional subspaces of \mathbb{C}^n , we get

$$\begin{aligned} (d_{\text{sp}}(\text{span}(\mathbf{u}_{J^c}), \text{span}(\tilde{\mathbf{u}}_{\hat{J}^c})))^2 &\leq \frac{1}{n - q} \sum_{j \in J^c} \frac{\|(\tilde{M} - M)\mathbf{u}_j\|_2^2}{(\min_{k \in J} |\lambda_k - \lambda_j| - \|\tilde{M} - M\|_2)^2} \\ &\leq \frac{\|\tilde{M} - M\|_2^2}{(\text{sep}(\lambda_J, \lambda_{J^c}) - \|\tilde{M} - M\|_2)^2}. \end{aligned}$$

But since $\text{span}(\mathbf{u}_{J^c})$ and $\text{span}(\tilde{\mathbf{u}}_{\tilde{J}^c})$ are orthogonal complements of $\text{span}(\mathbf{u}_J)$ and $\text{span}(\tilde{\mathbf{u}}_{\tilde{J}})$ respectively, from Lemma 2 we have $(n - q)(d_{\text{sp}}(\text{span}(\mathbf{u}_{J^c}), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}^c})))^2 = q(d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})))^2$. This gives us from the above

$$(d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})))^2 \leq \frac{1}{q} \sum_{j \in J^c} \frac{\|(\tilde{M} - M)\mathbf{u}_j\|_2^2}{(\min_{k \in J} |\lambda_k - \lambda_j| - \|\tilde{M} - M\|_2)^2} \quad (42)$$

$$\leq \frac{n - q}{q} \frac{\|\tilde{M} - M\|_2^2}{(\text{sep}(\lambda_J, \lambda_{J^c}) - \|\tilde{M} - M\|_2)^2}. \quad (43)$$

Combining (41) and (43) gives the first result of the proposition.

The second result can be obtained directly using Corollary 5.2 and observing that due to (39), $\text{sep}(\lambda_J, \tilde{\lambda}_{\tilde{J}^c}) \geq \text{sep}(\lambda_J, \lambda_{J^c}) - \|\tilde{M} - M\|_2$ (and analogously $\text{sep}(\lambda_{J^c}, \tilde{\lambda}_{\tilde{J}}) \geq \text{sep}(\lambda_J, \lambda_{J^c}) - \|\tilde{M} - M\|_2$). \square

Assuming $q \leq n/2$, it is worth noting that defining $\epsilon = \frac{1}{2} \text{sep}(\lambda_J, \lambda_{J^c}) - \|\tilde{M} - M\|_2$, the second inequality of the first result in the above proposition becomes $d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) \leq \frac{\|\tilde{M} - M\|_2}{\|\tilde{M} - M\|_2 + 2\epsilon}$. Thus, with $\epsilon \rightarrow 0$, this inequality becomes $d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) < 1$, rendering the result uninformative /redundant. Thus, the higher the separation between λ_J and λ_{J^c} (relative to $\|\tilde{M} - M\|_2$), the tighter will be the upper bound in the result of the proposition.

An interpretation of the result in the above proposition is that a perturbation $\tilde{M} - M$ of the matrix M will result in a perturbation in the invariant subspace $\text{span}(\mathbf{u}_J)$ such that the distance between the subspace and its perturbed counterpart is bounded by the upper bounds mentioned in the proposition. One key feature of the proposition, however, is that the upper bound in the inequality does not depend on \tilde{J} . As a consequence, for any other size- q subset \tilde{J} of N such that $\text{span}(\mathbf{u}_{\tilde{J}})$ is closer to $\text{span}(\mathbf{u}_J)$ than $\text{span}(\mathbf{u}_{\tilde{J}})$ still satisfies the same upper bound. That is, if $\|\tilde{M} - M\|_2 < \frac{1}{2} \text{sep}(\lambda_J, \lambda_{J^c})$, then

$$\begin{aligned} \min_{\tilde{J} \in S_{q,n}} d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) &\leq \min \left(1, \sqrt{\frac{n - q}{q}} \frac{\|\tilde{M} - M\|_2}{\text{sep}(\lambda_J, \lambda_{J^c}) - \|\tilde{M} - M\|_2} \right) \\ \min_{\tilde{J} \in S_{q,n}} d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) &\leq \frac{\frac{1}{\sqrt{2q}} \|\tilde{M} - M\|_F}{\text{sep}(\lambda_J, \lambda_{J^c}) - \|\tilde{M} - M\|_2}, \end{aligned} \quad (44)$$

where $S_{q,n}$ is the set of all q -element subsets of $N = \{1, 2, \dots, n\}$.

4 Application to null-space perturbation in the context of a graph connection problem

We consider a simple application of the above results in the context of a graph theory problem. Some definitions and basic properties of a weighted, undirected, simple graphs are listed below [10].

- 1 A graph G consists of a set of n vertices $\mathcal{V}(G) = \{v_1, v_2, \dots, v_n\}$ and an edge set $\mathcal{E}(G) \subseteq \mathcal{V}(G) \times_{\text{sym}} \mathcal{V}(G)$ (where ‘ \times_{sym} ’ represent the symmetric Cartesian product so that for the undirected graph the order of the vertices in an edge is irrelevant, making $(v_k, v_l) = (v_l, v_k)$). Each edge $(v_k, v_l) \in \mathcal{E}(G)$ is assigned a positive real weight $A_{kl}(= A_{lk})$. Nonexistent edges are implicitly assumed to have zero edge weight so that $A_{kl} = 0, \forall (v_k, v_l) \notin \mathcal{E}(G)$. The matrix $A \in \mathbb{R}^{n \times n}$ is called the *weighted adjacency*

matrix of the graph G and is a symmetric matrix with zero diagonal for an undirected, simple graph.

- 2 The *weighted degree matrix* D is an $n \times n$ diagonal matrix in which the k th diagonal element is the sum of the elements in the k th row (equivalently, k th column) of A . Thus D_{kk} is the sum of the weights of the edges emanating from v_k (also called the *degree* of the vertex).
- 3 The *weighted Laplacian matrix* of the graph is defined as $L = D - A$. An eigenvector of L is an n -dimensional real vector and can be interpreted as a distribution over the vertices (with the k th element of the vector being the value associated to $v_k \in \mathcal{E}(G)$).
- 4 The eigenvalues of L are nonnegative. The null-space of L for a graph with q disjoint components is q -dimensional, with the null-space spanned by vectors corresponding to distributions that are uniform over the vertices of each of those components. Without loss of generality we index the eigenvalues in an increasing order of their magnitudes so that $0 = \lambda_1 = \lambda_2 = \dots = \lambda_q \leq \lambda_{q+1} \leq \lambda_{q+2} \leq \dots \leq \lambda_n$. The corresponding unit eigenvectors are $\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_n$. Note that since a graph has at least one connected component, $\lambda_1 = 0$ for any graph. Furthermore, without loss of generality, we choose \mathbf{u}_j to be a distribution that is uniformly positive over the vertices of G_j and zero over the rest of the vertices in the graph.
- 5 Define $J = \{1, 2, \dots, q\}$, so that $\text{span}(\mathbf{u}_J)$ is the null-space of L .

If G has q disjoint components, we define $G_j, j = 1, 2, \dots, q$, to be the subgraph constituting of the vertices and edges in the j th component only. Thus, $\mathcal{V}(G) = \bigcup_{j=1}^q \mathcal{V}(G_j)$ and $\mathcal{E}(G) = \bigcup_{j=1}^q \mathcal{E}(G_j)$ (more compactly, we write $G = \bigcup_{j=1}^q G_j$). We also define the collection of these subgraphs as

$$\mathcal{G} = \{G_1, G_2, \dots, G_q\}.$$

We are interested in understanding perturbation of the invariant subspace $\text{span}(\mathbf{u}_J)$ (the null-space) of L as new edges are established between the different disjoint components (henceforth also referred to as “clusters”) of the graph. Let the graph constructed by establishing the inter-cluster edges be \tilde{G} with \tilde{A} , \tilde{D} , and \tilde{L} its adjacency, degree, and Laplacian matrices respectively. Note that since \tilde{G} is constructed by just adding edges between the subgraphs $\{G_j\}_{j=1,2,\dots,q}$ of G , all of these subgraphs are induced subgraphs of \tilde{G} .

4.1 Computation of $\|(\tilde{L} - L)\mathbf{u}_j\|_2$

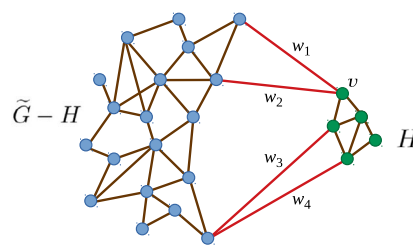
For any induced subgraph $H \in \tilde{G}$, we consider the edges that connect vertices in H to vertices not in H (*inter-cluster edges*). These are edges of the form (v_k, v_l) such that $v_k \in \mathcal{V}(H), v_l \notin \mathcal{V}(H)$. We define a few quantities involving the weights on such edges.

Definition 5

1. *External degree of a vertex relative to a subgraph*: Given a subgraph $H \subseteq \tilde{G}$ and a vertex $v_k \in \mathcal{V}(H)$, the *external degree of v_k relative to H in \tilde{G}* is defined as the sum of the weights on edges connecting v_k to vertices outside H :

$$\mathcal{ED}_{H,\tilde{G}}(v_k) = \sum_{\{l|v_l \notin \mathcal{V}(H)\}} \tilde{A}_{kl}. \quad (45)$$

Figure 5 An example graph \tilde{G} and induced subgraph H . Weight values on the inter-cluster edges are written symbolically. In this example, $\mathcal{ED}_{H,\tilde{G}}(v) = w_1 + w_2$, $\mathcal{CP}_{\tilde{G}}(H) = \frac{1}{5}(((w_1 + w_2)^2 + w_3^2 + w_4^2) + (w_1^2 + w_2^2 + (w_3 + w_4)^2))$, and $\mathcal{MED}_{\tilde{G}}(H) = \max(w_1 + w_2, w_3, w_4)$



2. *Coupling of a subgraph in a graph*: Given an induced subgraph $H \subseteq \tilde{G}$, we define the *coupling of H in \tilde{G}* as

$$\begin{aligned} \mathcal{CP}_{\tilde{G}}(H) &= \frac{1}{|\mathcal{V}(H)|} \left(\sum_{\{k|v_k \in \mathcal{V}(H)\}} (\mathcal{ED}_{H,\tilde{G}}(v_k))^2 + \sum_{\{l|v_l \notin \mathcal{V}(H)\}} (\mathcal{ED}_{(\tilde{G}-H),\tilde{G}}(v_l))^2 \right) \\ &= \frac{1}{|\mathcal{V}(H)|} \left(\sum_{\{k|v_k \in \mathcal{V}(H)\}} \left(\sum_{\{l|v_l \notin \mathcal{V}(H)\}} \tilde{A}_{kl} \right)^2 + \sum_{\{l|v_l \notin \mathcal{V}(H)\}} \left(\sum_{\{k|v_k \in \mathcal{V}(H)\}} \tilde{A}_{kl} \right)^2 \right), \end{aligned} \quad (46)$$

where $(\tilde{G} - H)$ is the induced subgraph of \tilde{G} constituting of all the vertices not in H . That is, $\mathcal{V}(\tilde{G} - H) = \{v \in \mathcal{V}(\tilde{G}) | v \notin \mathcal{V}(H)\}$ and $\mathcal{E}(\tilde{G} - H) = \{(v, w) \in \mathcal{E}(\tilde{G}) | v, w \notin \mathcal{V}(H)\}$.

3. *Maximum external degree of vertices in a subgraph*: Given a subgraph $H \subseteq \tilde{G}$, the *maximum external degree of vertices in H in \tilde{G}* is defined as the maximum value of the external degrees of vertices in H relative to H in \tilde{G} :

$$\mathcal{MED}_{\tilde{G}}(H) = \max_{v \in \mathcal{V}(H)} \mathcal{ED}_{H,\tilde{G}}(v) = \max_{\{k|v_k \in \mathcal{V}(H)\}} \sum_{\{l|v_l \notin \mathcal{V}(H)\}} \tilde{A}_{kl}. \quad (47)$$

Note that the computation of the above quantities requires the knowledge of only the weights on edges connecting vertices in H to vertices outside H in \tilde{G} (see Fig. 5 for an example).

In the definition of $\mathcal{CP}_{\tilde{G}}$, referring to H as a *cluster* and considering the rest of the graph another cluster, the quantity within the innermost brackets is the sum of the weights on inter-cluster edges connected to a vertex, which is squared and summed over all the vertices that have at least one inter-cluster edge connected to it. This quantity is then divided by the number of vertices in H . Thus a large subgraph which is weakly connected to the rest of the graph will have a lower coupling value.

The following lemma provides bounds on $\mathcal{CP}_{\tilde{G}}(H)$ in terms of a simpler summation over the inter-cluster edge weights (or square thereof).

Lemma 8

$$\frac{2}{|\mathcal{V}(H)|} \sum_{\substack{\{k,l|v_k \in \mathcal{V}(H), \\ v_l \notin \mathcal{V}(H)\}}} \tilde{A}_{kl}^2 \leq \mathcal{CP}_{\tilde{G}}(H) \leq \frac{2}{|\mathcal{V}(H)|} \left(\sum_{\substack{\{k,l|v_k \in \mathcal{V}(H), \\ v_l \notin \mathcal{V}(H)\}}} \tilde{A}_{kl} \right)^2. \quad (48)$$

Proof The proof follows directly using the fact that for a set of positive numbers $\alpha_h, h \in S$, $\sum_{h \in S} \alpha_h^2 \leq (\sum_{h \in S} \alpha_h)^2$. \square

Notations and assumptions for the rest of the paper In the rest of the paper we assume that G is a graph with q disjoint components $G = \{G_1, G_2, \dots, G_q\}$ and \tilde{G} is the graph obtained by establishing edges between the components (so that each G_j is an induced subgraph of both G and \tilde{G}). The Laplacian matrices of the two graphs are L and \tilde{L} respectively. Since G has q connected components, its null-space is q dimensional (with corresponding eigenvalues $\lambda_1 = \lambda_2 = \dots = \lambda_q$), for which we choose a basis $\{\mathbf{u}_j\}_{j=1,2,\dots,q}$ such that the distribution corresponding to \mathbf{u}_j is uniform and positive on the vertices in G_j and zero everywhere else.

A weaker version of the following lemma appears in the author's prior work [11, 12] and expresses the quantity $\|(\tilde{L} - L)\mathbf{u}_j\|_2$ in terms of the weights on edges connecting vertices in G_j to vertices outside G_j in \tilde{G} .

Lemma 9 For all $j \in \{1, 2, \dots, q\}$,

$$\|(\tilde{L} - L)\mathbf{u}_j\|_2^2 = \mathcal{CP}_{\tilde{G}}(G_j). \quad (49)$$

Proof Suppose $v_k \in \mathcal{V}(G_j) \subseteq \mathcal{V}(G)$. Since \tilde{D}_{kk} and D_{kk} are the degrees of the vertex in the graphs \tilde{G} and G respectively, they are equal iff all the neighbors of v_k are in G_j . Otherwise $\tilde{D}_{kk} - D_{kk}$ is the net outgoing degree of the vertex v_k from the subgraph G_j . That is, if $v_k \in \mathcal{V}(G_j)$, then

$$\tilde{D}_{kk} - D_{kk} = \sum_{\{l | v_l \notin \mathcal{V}(G_j)\}} \tilde{A}_{kl}. \quad (50)$$

An edge (v_k, v_l) exists in both \tilde{G} and G (and have the same weight, i.e., $\tilde{A}_{kl} = A_{kl}$) iff v_k and v_l belong to the same subgraph G_j . Otherwise $A_{kl} = 0$ (the edge is nonexistent in G). Thus,

$$\tilde{A}_{kl} - A_{kl} = \begin{cases} \tilde{A}_{kl}, & \text{if } v_k \in \mathcal{V}(G_j), v_l \notin \mathcal{V}(G_j), \\ 0, & \text{otherwise.} \end{cases} \quad (51)$$

Next we consider the vector \mathbf{u}_j (for $j = 1, \dots, q$), which by definition is nonzero and uniform only on vertices in the subgraph G_j . Let u_{lj} be the l th element of the unit vector \mathbf{u}_j . Since $|\mathcal{V}(G_j)|$ of the elements of the vector are nonzero and uniform, we have

$$u_{lj} = \begin{cases} \frac{1}{\sqrt{|\mathcal{V}(G_j)|}}, & \text{if } v_l \in \mathcal{V}(G_j), \\ 0, & \text{otherwise.} \end{cases} \quad (52)$$

Thus the k th element of the vector $(\tilde{L} - L)\mathbf{u}_j$,

$$\begin{aligned} [(\tilde{L} - L)\mathbf{u}_j]_k &= \sum_l (\tilde{D}_{kl} - \tilde{A}_{kl} - D_{kl} + A_{kl})u_{lj} \\ &= (\tilde{D}_{kk} - D_{kk})u_{kj} - \sum_l (\tilde{A}_{kl} - A_{kl})u_{lj} \end{aligned}$$

$$\begin{aligned}
& \text{(since } \tilde{D} \text{ and } D \text{ are diagonal matrices)} \\
&= \left(\frac{1}{\sqrt{|\mathcal{V}(G_j)|}} \begin{cases} (\tilde{D}_{kk} - D_{kk}), & \text{if } v_k \in \mathcal{V}(G_j) \\ 0, & \text{otherwise} \end{cases} \right) \\
&\quad - \left(\frac{1}{\sqrt{|\mathcal{V}(G_j)|}} \sum_{\{l|v_l \in \mathcal{V}(G_j)\}} (\tilde{A}_{kl} - A_{kl}) \right) \quad \text{(using (52))} \\
&= \frac{1}{\sqrt{|\mathcal{V}(G_j)|}} \left(\begin{cases} \sum_{\{l|v_l \notin \mathcal{V}(G_j)\}} \tilde{A}_{kl}, & \text{if } v_k \in \mathcal{V}(G_j) \\ 0, & \text{otherwise} \end{cases} \right. \\
&\quad \left. - \sum_{\{l|v_l \in \mathcal{V}(G_j)\}} \begin{cases} \tilde{A}_{kl}, & \text{if } v_k \notin \mathcal{V}(G_j) \\ 0, & \text{otherwise} \end{cases} \right) \quad \text{(using (50) and (51))} \\
&= \frac{1}{\sqrt{|\mathcal{V}(G_j)|}} \begin{cases} \sum_{\{l|v_l \notin \mathcal{V}(G_j)\}} \tilde{A}_{kl}, & \text{if } v_k \in \mathcal{V}(G_j) \\ - \sum_{\{l|v_l \in \mathcal{V}(G_j)\}} \tilde{A}_{kl}, & \text{if } v_k \notin \mathcal{V}(G_j). \end{cases} \quad (53)
\end{aligned}$$

Thus,

$$\begin{aligned}
& \|(\tilde{L} - L)\mathbf{u}_j\|_2^2 \\
&= \frac{1}{|\mathcal{V}(G_j)|} \left(\sum_{\{k|v_k \in \mathcal{V}(G_j)\}} \left(\sum_{\{l|v_l \notin \mathcal{V}(G_j)\}} \tilde{A}_{kl} \right)^2 + \sum_{\{k|v_k \notin \mathcal{V}(G_j)\}} \left(\sum_{\{l|v_l \in \mathcal{V}(G_j)\}} \tilde{A}_{kl} \right)^2 \right). \quad (54)
\end{aligned}$$

□

Lemma 10

$$\|\tilde{L} - L\|_2 \leq 2 \max_{j \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_j). \quad (55)$$

Proof Suppose $v_k \in \mathcal{V}(G_{j(k)})$ (where $j : \{1, 2, \dots, |\mathcal{V}(G)|\} \rightarrow \{1, 2, \dots, q\}$ maps the index of a vertex to the index of the subgraph in \mathcal{G} that the vertex belongs to). The sum of the elements of the k th row of $(\tilde{A} - A)$ is

$$\begin{aligned}
\sum_l (\tilde{A}_{kl} - A_{kl}) &= \sum_l \begin{cases} \tilde{A}_{kl}, & \text{if } v_l \notin \mathcal{V}(G_{j(k)}) \\ 0, & \text{otherwise} \end{cases} \quad \text{(using (51))} \\
&= \sum_{\{l|v_l \notin \mathcal{V}(G_{j(k)})\}} \tilde{A}_{kl} \\
&= \mathcal{ED}_{G_{j(k)}, \tilde{G}}(v_k) \quad \text{(Definition 5)}. \quad (56)
\end{aligned}$$

Since $(\tilde{A} - A)$ is a symmetric matrix, its 2-norm is equal to its spectral radius $\rho(\tilde{A} - A)$. Furthermore, since all elements of $(\tilde{A} - A)$ are nonnegative, using the Perron–Frobenius theorem [13], we get

$$\begin{aligned}
\|\tilde{A} - A\|_2 &= \rho(\tilde{A} - A) \leq \max_{k \in N} \mathcal{ED}_{G_{j(k)}, \tilde{G}}(v_k) \\
&= \max_{j \in \{1, \dots, q\}} \max_{\substack{\{k\} \\ v_k \in \mathcal{V}(G_j)}} \mathcal{ED}_{G_{j(k)}, \tilde{G}}(v_k)
\end{aligned}$$

$$\begin{aligned}
 & \text{(since maximizing over all vertices in } \tilde{G} \text{ is same as} \\
 & \text{maximizing over the subgraphs } G_j \text{ and for each} \\
 & \text{subgraph maximizing over the vertices in the subgraph)} \\
 & = \max_{j \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_j). \tag{57}
 \end{aligned}$$

Again, since $(\tilde{D} - D)$ is a diagonal matrix with positive diagonal elements (due to (50)), its 2-norm is the maximum out of its diagonal elements. That is,

$$\begin{aligned}
 \|\tilde{D} - D\|_2 &= \max_{k \in N} (\tilde{D}_{kk} - D_{kk}) = \max_{k \in N} \sum_{\{l | v_l \notin \mathcal{V}(G_{j(k)})\}} \tilde{A}_{kl} \quad (\text{using (50)}) \\
 &= \max_{k \in N} \mathcal{ED}_{G_{j(k)}, \tilde{G}}(v_k) \\
 &= \max_{j \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_j) \quad (\text{following similar steps as in (56) and (57)}). \tag{58}
 \end{aligned}$$

Thus,

$$\begin{aligned}
 \|\tilde{L} - L\|_2 &= \|(\tilde{D} - D) - (\tilde{A} - A)\|_2 \leq \|\tilde{D} - D\|_2 + \|\tilde{A} - A\|_2 \\
 &\leq 2 \max_{j \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_j). \quad \square
 \end{aligned}$$

In the following discussions, without loss of generality, we assume that the eigenvalues of \tilde{L} are indexed in an increasing order of magnitude $0 = \tilde{\lambda}_1 \leq \tilde{\lambda}_2 \leq \dots \leq \tilde{\lambda}_n$. The corresponding eigenvectors are $\tilde{\mathbf{u}}_1, \tilde{\mathbf{u}}_2, \dots, \tilde{\mathbf{u}}_n$.

4.2 Bounds on null-space perturbation with known spectrum of \tilde{L}

The following proposition gives a bound on the perturbation of the null-space of L upon introducing edges between the subgraphs in $G = \{G_1, G_2, \dots, G_q\}$ by considering the sub-space distance between the null-space of L and a specific invariant sub-space of \tilde{L} .

Proposition 3 Choose $\tilde{J} = \{1, 2, \dots, q\}$. Then

$$d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\tilde{J}})) \leq \frac{1}{\tilde{\lambda}_{q+1}} \sqrt{\frac{1}{q} \sum_{j=1}^q \mathcal{CP}_{\tilde{G}}(G_j)}. \tag{59}$$

Proof We first note that due to Lemma 9 $\sqrt{\mathcal{CP}_{\tilde{G}}(G_j)} = \|(\tilde{L} - L)\mathbf{u}_j\|_2, \forall j \in \{1, 2, \dots, q\}$. The proof then follows from Proposition 1 by setting $\kappa_j = 0, \forall j = 1, 2, \dots, q$ and noting that $\min_{j \in \{1, 2, \dots, q\}, j' \in \{q+1, q+2, \dots, n\}} |\tilde{\lambda}_{j'} - \lambda_j| = \tilde{\lambda}_{q+1}$. \square

The results of Proposition 3 can be re-interpreted by considering G to be the graph obtained by *cutting* \tilde{G} into q -subgraphs. We call the set of subgraphs hence constructed upon performing the cut $G = \{G_1, G_2, \dots, G_q\}$ a q -cut of \tilde{G} . Given a graph \tilde{G} , we consider all possible q -cuts of \tilde{G} . A q -cut $G = \{G_1, G_2, \dots, G_q\}$ results in a graph $G = \bigcup_{j=1}^q G_j$ with q disjoint components. The following corollary is then a direct consequence of the proposition.

Corollary 6 Given a graph \tilde{G} (with Laplacian \tilde{L} with eigenvalues $0 = \tilde{\lambda}_1 \leq \dots \leq \tilde{\lambda}_n$ and corresponding eigenvectors $\tilde{\mathbf{u}}_1, \dots, \tilde{\mathbf{u}}_n$), let \mathcal{G} be the set of all q -cuts of \tilde{G} . We consider a q -cut such that the sum of the couplings of the resultant q subgraphs in \tilde{G} is minimum. That is,

$$G^* \in \arg \min_{G \in \mathcal{G}} \sum_{G' \in G} \mathcal{CP}_{\tilde{G}}(G'). \quad (60)$$

Let the corresponding graph $G^* = \bigcup_{G' \in G^*} G'$ have eigenvalues $0 = \lambda_1^* = \lambda_2^* = \dots = \lambda_q^* \leq \lambda_{q+1}^* \leq \dots \leq \lambda_n^*$ and corresponding eigenvectors $\mathbf{u}_1^*, \mathbf{u}_2^*, \dots, \mathbf{u}_n^*$. Then

$$d_{\text{sp}}(\text{span}(\{\tilde{\mathbf{u}}_1, \dots, \tilde{\mathbf{u}}_q\}), \text{span}(\{\mathbf{u}_1^*, \dots, \mathbf{u}_q^*\})) \leq \frac{1}{\tilde{\lambda}_{q+1}} \sqrt{\frac{1}{q} \sum_{G' \in G^*} \mathcal{CP}_{\tilde{G}}(G')}. \quad (61)$$

The interpretation of the above corollary is that the “best” q -cut of a graph \tilde{G} (minimizing total inter-cluster coupling, as defined by (60)) results in a graph such that the distance between the nullspace of the cut graph’s Laplacian and the space spanned by the first q eigenvectors of the Laplacian of \tilde{G} is bounded above by a quantity proportional to the total inter-cluster coupling (which was minimized in the first place).

4.3 Bounds on null-space perturbation with known spectrum of L

Proposition 4 If $\max_{j \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_j) < \frac{\lambda_{q+1}}{4}$, then

$$\begin{aligned} d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\hat{J}})) &\leq \frac{\sqrt{\frac{1}{q} \sum_{j=1}^q \mathcal{CP}_{\tilde{G}}(G_j)}}{\lambda_{q+1} - 2 \max_{k \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_k)} \\ &\leq \frac{2 \max_{j \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_j)}{\lambda_{q+1} - 2 \max_{j \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_j)}, \end{aligned} \quad (62)$$

where $\hat{J} = \{1, 2, \dots, q\} = \{j' \mid \min_{j \in N} |\tilde{\lambda}_{j'} - \lambda_j| = \tilde{\lambda}_{j'}\}$.

Proof Recall that the eigenvalues of the Laplacian L of G are $(0 =) \lambda_1 = \lambda_2 = \dots = \lambda_q \leq \lambda_{q+1} \leq \dots \leq \lambda_n$. Let $J = \{1, 2, \dots, q\}$ so that $J^c = \{q+1, q+2, \dots, n\}$ and $\text{sep}(\lambda_J, \lambda_{J^c}) = \lambda_{q+1}$.

Using Lemma 10, we have

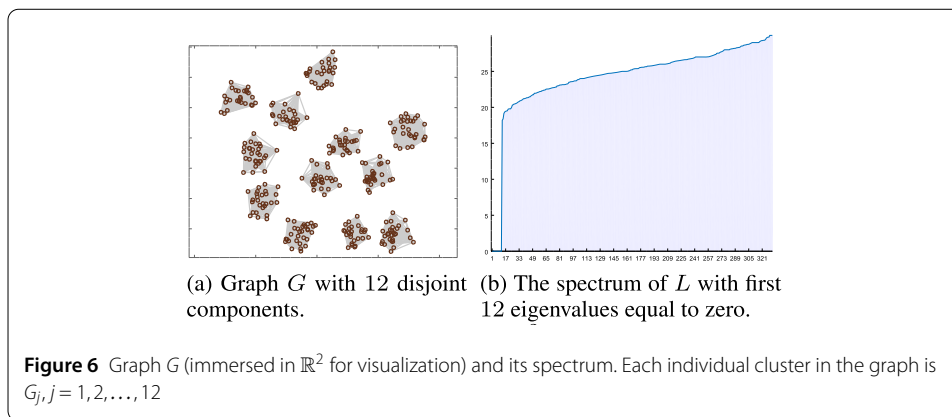
$$\|\tilde{L} - L\|_2 \leq 2 \max_{j \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_j) < \frac{\lambda_{q+1}}{2} = \frac{\text{sep}(\lambda_J, \lambda_{J^c})}{2}. \quad (63)$$

Thus the conditions for Lemma 7 and Proposition 2 hold, and \tilde{L} is a separation preserving perturbation of L . Hence, by Lemma 7, there exists a separation preserving partition $\{\tilde{\lambda}_{\hat{J}}, \tilde{\lambda}_{\hat{J}^c}\}$ of $\tilde{\lambda}_N$ such that

$$\hat{J} = \left\{ j' \mid \min_{j \in N} |\tilde{\lambda}_{j'} - \lambda_j| = \min_{j \in J} |\tilde{\lambda}_{j'} - \lambda_j| = \tilde{\lambda}_{j'} \right\} \quad (\text{since } \lambda_j = 0, \forall j \in J).$$

Thus, for any $j' \in \hat{J}$,

$$\begin{aligned} \tilde{\lambda}_{j'} &= \min_{j \in N} |\tilde{\lambda}_{j'} - \lambda_j| \leq \|\tilde{L} - L\|_2 \quad (\text{due to Corollary 3}) \\ &\leq \frac{\lambda_{q+1}}{2} \quad (\text{from (63)}). \end{aligned} \quad (64)$$



This implies that the elements of $\tilde{\lambda}_{\hat{J}}$ are closer to $0 (= \lambda_1 = \lambda_2 = \dots = \lambda_q)$ than they are to λ_{q+1} . Since \hat{J} has q -elements (due to Lemma 7.2) and is a unique set (by definition), we have $\tilde{\lambda}_{\hat{J}} = \{\tilde{\lambda}_1, \tilde{\lambda}_2, \dots, \tilde{\lambda}_q\}$ to be the set constituting of the lowest q eigenvalues of \tilde{L} . Thus, $\hat{J} = \{1, 2, \dots, q\}$.

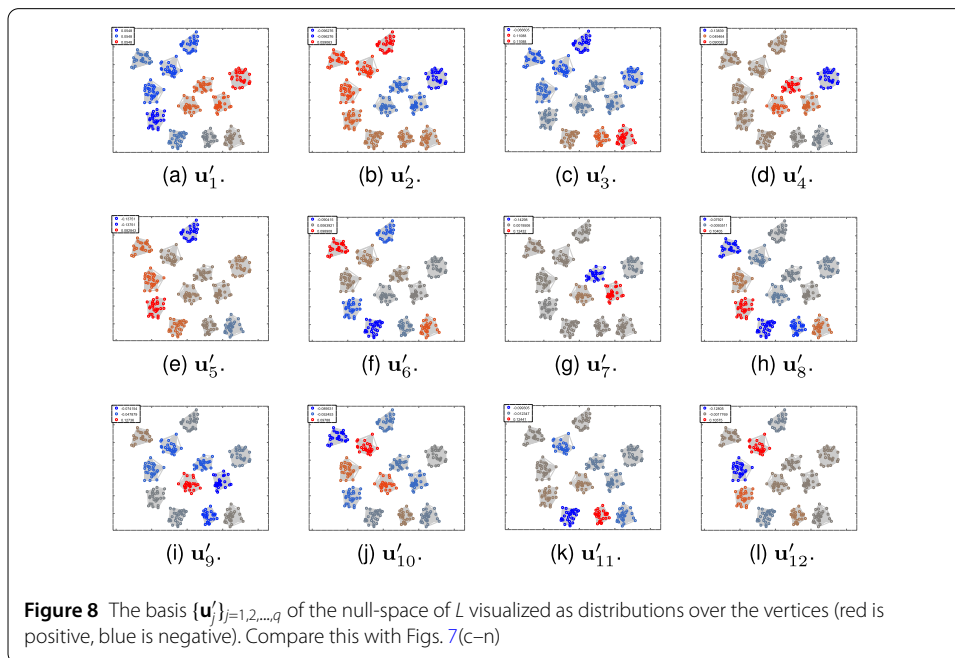
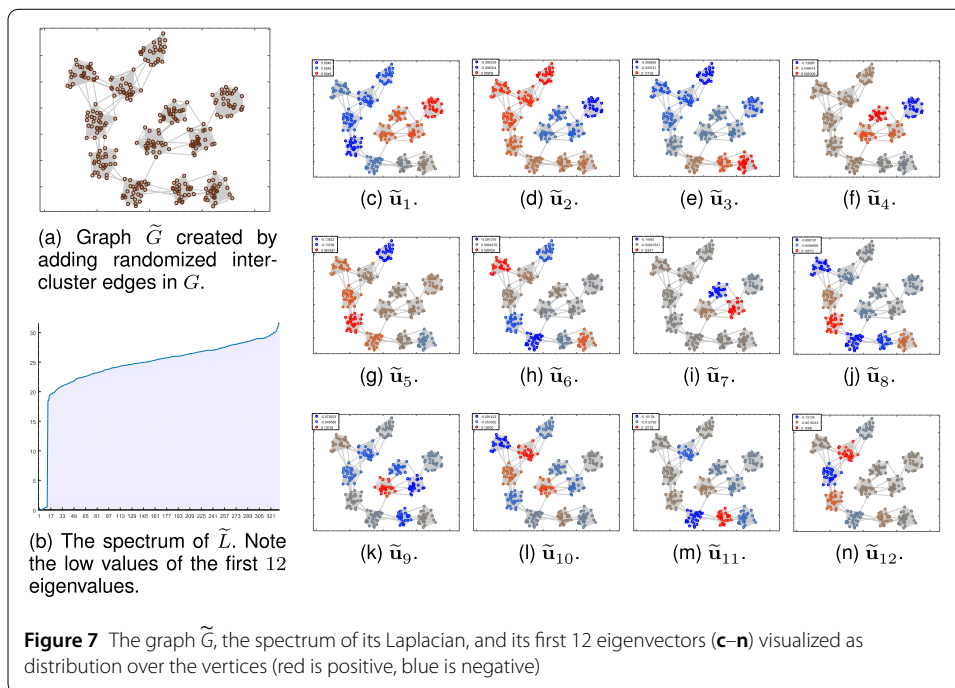
Since we showed that $\|\tilde{L} - L\|_2 \leq \frac{1}{2} \text{sep}(\lambda_j, \lambda_{j^c})$, as direct consequence of Proposition 2, we have the following:

$$\begin{aligned}
 & (d_{\text{sp}}(\text{span}(\mathbf{u}_J), \text{span}(\tilde{\mathbf{u}}_{\hat{J}})))^2 \\
 & \leq \frac{1}{q} \sum_{j \in J} \left(\frac{\|(\tilde{L} - L)\mathbf{u}_j\|_2}{\min_{k \in J^c} |\lambda_k - \lambda_j| - \|\tilde{L} - L\|_2} \right)^2 \\
 & \leq \frac{\frac{1}{q} \sum_{j \in J} \mathcal{CP}_{\tilde{G}}(G_j)}{(\lambda_{q+1} - 2 \max_{k \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_k))^2} \\
 & \quad \left(\text{using Lemma 9 and Lemma 10 and the} \right. \\
 & \quad \left. \text{fact that } \min_{k \in J^c} |\lambda_k - \lambda_j| = \lambda_{q+1}, \forall j \in \{1, 2, \dots, n\} \right) \\
 & \leq \left(\frac{2 \max_{j \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_j)}{\lambda_{q+1} - 2 \max_{j \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_j)} \right)^2 \\
 & \quad \left(\text{using Lemma 9 and 10, } \sum_{j \in J} \mathcal{CP}_{\tilde{G}}(G_j) = \sum_{j \in J} \|(\tilde{L} - L)\mathbf{u}_j\|_2^2 \right) \\
 & \leq q \|\tilde{L} - L\|_2^2 \leq q \left(2 \max_{j \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_j) \right)^2. \quad \square
 \end{aligned}$$

4.4 Example

As an illustration, we consider the graph G shown in Fig. 6 with 12 disjoint components, thus $q = 12$. The graph is generated with $n = 333$ vertices clustered into 12 components in a randomized manner with only intra-cluster edges. The weight on every edge is chosen to be 1. Figure 6(a) shows an immersion of the graph in \mathbb{R}^2 just for the purpose of visualization (the exact coordinates of the vertices have no significance).

We then construct \tilde{G} by establishing randomized edges between the components of G . The weight on every inter-cluster edge is also chosen to be 1. Figure 7(a) shows the immersion of the resultant graph.



Direct computation reveals that for these graphs, $\tilde{\lambda}_{q+1} = 18.436$ and $\frac{1}{q} \sum_{j=1}^q \mathcal{CP}_{\tilde{G}}(G_j) = 0.5417$. The L.H.S. of (59) is $d_{\text{sp}}(\text{span}(\mathbf{u}_{1,2,\dots,12}), \text{span}(\tilde{\mathbf{u}}_{1,2,\dots,12})) = 2.516 \times 10^{-2}$, while the R.H.S. is $\frac{\sqrt{\frac{1}{q} \sum_{j=1}^q \mathcal{CP}_{\tilde{G}}(G_j)}}{\tilde{\lambda}_{q+1}} = 3.992 \times 10^{-2}$, thus validating the result of Proposition 3.

Again, $\max_{j \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_j) = 3$ and $\frac{\lambda_{q+1}}{4} = 4.6091$, thus satisfying the condition for Proposition 4. The R.H.S. in (62) is $\frac{\sqrt{\frac{1}{q} \sum_{j=1}^q \mathcal{CP}_{\tilde{G}}(G_j)}}{\lambda_{q+1} - 2 \max_{k \in \{1, \dots, q\}} \mathcal{MED}_{\tilde{G}}(G_k)} = 6.036 \times 10^{-2}$, thus validating the result of the proposition.

Since the chosen basis $\{\mathbf{u}_j\}_{j=1,2,\dots,q}$ for the null-space of L consists of distributions such that \mathbf{u}_j is uniform and positive over vertices of G_j and zero everywhere else, this basis is not ideal for a visual comparison with $\{\tilde{\mathbf{u}}_j\}_{j=1,2,\dots,q}$. For a visual comparison between $\text{span}(\mathbf{u}_{\{1,2,\dots,q\}})$ and $\text{span}(\tilde{\mathbf{u}}_{\{1,2,\dots,q\}})$, we choose a basis for the null-space of L that is closest to $\{\tilde{\mathbf{u}}_j\}_{j=1,2,\dots,q}$: Define the $q \times q$ matrix $R = ([\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_q])^+ [\tilde{\mathbf{u}}_1, \tilde{\mathbf{u}}_2, \dots, \tilde{\mathbf{u}}_q]$, where $(\cdot)^+$ indicates the Moore–Penrose pseudoinverse. We need to choose a unitary matrix that is close to R . This is given by taking the SVD of $R = V \Sigma W^\dagger$ and defining $R' = V W^\dagger$. Then a basis for $\text{span}(\mathbf{u}_{\{1,2,\dots,q\}})$ is defined by the columns of $[\mathbf{u}_1, \mathbf{u}_2, \dots, \mathbf{u}_q] R' =: [\mathbf{u}'_1, \mathbf{u}'_2, \dots, \mathbf{u}'_q]$. Figure 8 shows these vectors as distributions over the vertices of G .

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Competing interests

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References

1. Davis, C., Kahan, W.M.: The rotation of eigenvectors by a perturbation. III. *SIAM J. Numer. Anal.* **7**(1), 1–46 (1970). <https://doi.org/10.1137/0707001>
2. Bhatia, R.: *Matrix Analysis*. Graduate Texts in Mathematics. Springer, Berlin (1996)
3. Stewart, G.W.: Error and perturbation bounds for subspaces associated with certain eigenvalue problems. *SIAM Rev.* **15**(4), 727–764 (1973)
4. Stewart, G.W., Sun, J.-g.: *Matrix Perturbation Theory*. Academic Press, San Diego (1990)
5. Bhattacharya, S.: On some bounds on the perturbation of invariant subspaces of normal matrices with application to a graph connection problem (2021). Preprint at [arXiv:2103.09413](https://arxiv.org/abs/2103.09413)
6. Andruchow, E.: Operators which are the difference of two projections. *J. Math. Anal. Appl.* **420**(2), 1634–1653 (2014). <https://doi.org/10.1016/j.jmaa.2014.06.022>
7. Golub, G.H., Van Loan, C.F., Van Loan, C.F., Van Loan, P.C.F.: *Matrix Computations*. Johns Hopkins Studies in the Mathematical Sciences. Johns Hopkins University Press, Baltimore (1996)
8. Damle, A., Sun, Y.: Uniform bounds for invariant subspace perturbations (2020) [arXiv:1905.07865](https://arxiv.org/abs/1905.07865)
9. Bauer, F.L., Fike, C.T.: Norms and exclusion theorems. *Numer. Math.* **2**(1), 137–144 (1960)
10. Godsil, C., Royle, C.D.G., Royle, G.F.: *Algebraic Graph Theory*. Graduate Texts in Mathematics. Springer, Berlin (2001)
11. Zhang, L., Sadler, B.M., Blum, R.S., Bhattacharya, S.: Inter-cluster transmission control using graph modal barriers (2020). Preprint [arXiv:2010.04790](https://arxiv.org/abs/2010.04790) [cs.LG]
12. Zhang, L., Sadler, B.M., Blum, R.S., Bhattacharya, S.: Inter-cluster transmission control using graph modal barriers. In: *IEEE Transactions on Signal and Information Processing over Networks*, pp. 1–1 (2021). <https://doi.org/10.1109/TSIPN.2021.3071219>
13. Berman, A., Plemmons, R.J.: *Nonnegative Matrices in the Mathematical Sciences*. Society for Industrial and Applied Mathematics, Philadelphia (1994). <https://doi.org/10.1137/1.9781611971262>